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Running Head: Bee richness dilutes multiple viruses

Title: Pollinator community species richness dilutes prevalence of multiple viruses within multiple host species

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Abstract

Most pathogens are embedded in complex communities composed of multiple interacting hosts, but we are still learning how community-level factors, such as host diversity, abundance, and composition, contribute to pathogen spread for many host–pathogen systems. Evaluating relationships among multiple pathogens and hosts may clarify whether particular host or pathogen traits consistently drive links between community factors and pathogen prevalence. Pollinators are a good system to test how community composition influences pathogen spread because pollinator communities are extremely variable and contain several multi-host pathogens

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30 transmitted on shared floral resources. We conducted a field survey of four pollinator species to
31 test the prevalence of three RNA viruses (deformed wing virus, black queen cell virus, and
32 sacbrood virus) among pollinator communities with variable species richness, abundance, and
33 composition. All three viruses showed a similar pattern of prevalence among hosts. *Apis*
34 *mellifera* and *Bombus impatiens* had significantly higher viral prevalence than *Lasioglossum* spp.
35 and *Eucera pruinosa*. In each species, lower virus prevalence was most strongly linked with
36 greater pollinator community species richness. In contrast, pollinator abundance, species-specific
37 pollinator abundance, and community composition were not associated with virus prevalence.
38 Our results support a consistent dilution effect for multiple viruses and host species. Pollinators
39 in species-rich communities had lower viral prevalence than pollinators from species-poor
40 communities, when accounting for differences in pollinator abundance. Species-rich
41 communities likely had lower viral prevalence because species-rich communities contained more
42 native bee species likely to be poor viral hosts than species-poor communities, and all
43 communities contained the highly competent hosts *A. mellifera* and *B. impatiens*. Interestingly,
44 the strength of the dilution effect was not consistent among hosts. Instead, host species with low
45 viral prevalence exhibited weaker dilution effects compared to hosts with high viral prevalence.
46 Therefore, host species susceptibility and competence for each virus may contribute to variation
47 in the strength of dilution effects. This study expands biodiversity–disease studies to the
48 pollinator–virus system, finding consistent evidence of the dilution effect among multiple similar
49 pathogens that infect ‘replicate’ host communities.

50
51 **Key Words:** *Apis mellifera*, black queen cell virus, *Bombus*, community composition, deformed
52 wing virus, dilution effect, biodiversity–disease, multi-host pathogens, native bees, sacbrood
53 virus.

54 **Introduction**

55 Host–pathogen interactions occur within complex ecological communities composed of
56 multiple host species and multiple pathogens, which can influence patterns of transmission and
57 disease outcomes. Heterogeneity among host species in their likelihood of encountering,
58 becoming infected (i.e. susceptibility), and transmitting pathogens to other hosts (i.e.
59 competency) contribute to variation in pathogen transmission and prevalence among

60 communities (Fenton et al. 2015). Therefore, the biodiversity, relative abundance, and identity of
61 hosts present in a community may influence pathogen prevalence (LoGiudice et al. 2003,
62 Keesing et al. 2006). For example, differences in bird community diversity, relative abundance,
63 and composition predict differences in West Nile virus prevalence in birds and humans due to
64 heterogeneity in bird host competence and transmission rates (Ezenwa et al. 2006, Kilpatrick et
65 al. 2006).

66 Pathogen characteristics, such as host ranges and modes of transmission, also have strong
67 effects on patterns of multi-host pathogen prevalence (Woolhouse and Gowtage-Sequeria 2005).
68 Multiple pathogens often circulate among the same communities of hosts, but pathogens with
69 different traits are likely to show different relationships between biodiversity and infectious
70 disease prevalence (hereafter, ‘biodiversity–disease relationship’)(Rohr et al. 2020). For
71 example, Wood et al. found that pathogen characteristics were important for determining
72 whether greater wildlife biodiversity could reduce, increase, or not affect prevalence of many
73 human pathogens (Wood et al. 2014a). Thus far, few studies have evaluated variability among
74 hosts and pathogens in how host community factors, such as host diversity, abundance, and
75 composition, impact biodiversity–disease relationships.

76 Although the relationships between host communities and pathogen prevalence are not
77 simple, three community-level variables are thought to influence disease dynamics: host species
78 diversity, host abundance, and community composition (Keesing et al. 2010, Roche et al. 2012).
79 Greater host biodiversity is hypothesized to reduce pathogen prevalence through the ‘dilution
80 effect’ (Keesing et al. 2006). The dilution effect is predicted to occur when species-poor
81 communities are dominated by highly competent hosts, and additional species in diverse
82 communities are less competent hosts or reduce encounters, transmission, or density of the
83 competent hosts (Ostfeld and Keesing 2000, Keesing et al. 2006). The dilution effect is
84 supported by the tick-borne Lyme disease system. High vertebrate biodiversity reduces *Borrelia*
85 *burgdorferi* prevalence because ticks are more likely to feed on less competent hosts in diverse
86 communities compared to species-poor communities dominated by highly competent white-
87 footed mice (Ostfeld and Keesing 2000). Though there is growing evidence for the dilution
88 effect in many multi-host–pathogen systems (Ezenwa et al. 2006, Clay et al. 2009, Johnson et al.
89 2013b, Venesky et al. 2014), other studies have found different biodiversity–disease
90 relationships (Salkeld et al. 2013, Luis et al. 2018).

91 Biodiversity–disease relationships can also exhibit the ‘amplification effect’, where
92 greater host species diversity increases pathogen prevalence (Keesing et al. 2006). The
93 amplification effect is likely when highly competent hosts are found in species-rich rather than
94 species-poor communities, or additional species facilitate greater pathogen transmission among
95 hosts (Keesing et al. 2006, Luis et al. 2018). Additionally, some pathogens are not influenced by
96 changes in community diversity, and therefore could have a neutral biodiversity–disease
97 relationship (Wood et al. 2014a, Rohr et al. 2020). There is much interest in when different
98 biodiversity–disease relationships are observed and their underlying mechanisms (Randolph and
99 Dobson 2012, Wood and Lafferty 2013, Rohr et al. 2020). Expanding biodiversity–disease
100 studies to additional multi-host–pathogen systems is an important frontier to further understand
101 the conditions at the community-level that lead to dilution, amplification, or neutral effects.

102 A central challenge in empirical biodiversity–disease studies revolves around
103 disentangling the effects of host diversity, host abundance, and host identity (i.e. community
104 composition) on pathogen prevalence to understand the mechanisms that drive biodiversity–
105 disease relationships. Host abundance scales with species richness in most natural communities
106 (Mihaljevic et al. 2014), therefore it is important to evaluate the relative contributions of host
107 diversity and host abundance to observed biodiversity–disease relationships to elucidate their
108 underlying mechanisms (Rudolf and Antonovics 2005). As biodiversity increases, the addition of
109 less competent hosts can reduce the abundance of highly competent hosts to subsequently reduce
110 pathogen transmission and prevalence, known as the ‘susceptible host regulation’ mechanism of
111 the dilution effect (Keesing et al. 2006). For example, Mitchell et al. (2002) found reduced
112 disease severity of several species-specific foliar fungal diseases in species-rich plant
113 communities, but the pattern was driven by lower species-specific densities in the species-rich
114 plots rather than biodiversity *per se*. Alternatively, diverse communities that contain multiple
115 competent host species could result in a greater abundance of susceptible hosts and maintain
116 higher levels of pathogen prevalence (i.e. amplification) (Holt et al. 2003). Therefore, it is
117 critical to control for host density in biodiversity–disease studies, especially for multi-host
118 pathogens that are shared among several abundant and susceptible host species in a community.

119 Host community composition, including both species identity and relative abundance, can
120 also have a strong effects on the relationship between host diversity and pathogen prevalence
121 (Randolph and Dobson 2012, Mihaljevic et al. 2014). Host species differ in many factors (e.g.

122 susceptibility, infectiousness, behavior, and competence), so the presence or absence of
123 particular host species can alter patterns of pathogen prevalence (Ostfeld and LoGiudice 2003,
124 Fenton et al. 2015). If highly competent hosts are common in species-poor communities and
125 additional species in diverse communities are more likely to be less competent hosts, then a
126 dilution effect pattern is more likely to occur. For example, Johnson et al. (2013b) found that
127 species-poor communities dominated by the highly-competent amphibian host *Pseudacris regilla*
128 tended to have higher infection prevalence for the trematode parasite *Ribeiroia ondatrae*
129 compared to more diverse communities composed of more pathogen-resistant species. In this
130 case, the dilution effect pattern is due to the presence of particular host species rather than host
131 species richness alone. Previous studies have shown that the presence of highly competent or low
132 competence “diluter” hosts can be important predictors of pathogen prevalence in diverse host–
133 pathogen systems, including Lyme disease in vertebrates (LoGiudice et al. 2003), West Nile
134 Virus in birds (Ezenwa et al. 2006), *Batrachochytrium dendrobatidis* in amphibians (Becker et
135 al. 2014, Venesky et al. 2014), and *Metschnikowia* fungus in *Daphnia* (Strauss et al. 2018).
136 Though many studies have tested the relative impacts of host community diversity, abundance,
137 and composition on pathogen prevalence, few studies have compared the effects these factors on
138 prevalence of several pathogens that infect the same sets of hosts (but see Johnson et al. 2013a).
139 Systems with multiple hosts and multiple pathogens provide a powerful model to test
140 which community-level factors influence pathogen transmission and prevalence because we can
141 tease apart commonalities among similar hosts or shared pathogens. Similar traits among hosts or
142 pathogens can lead to consistently negative biodiversity–disease relationships, where pathogen
143 prevalence is diluted by increased host diversity or other community factors (Ezenwa et al. 2006,
144 Johnson et al. 2013a, 2013b, Becker et al. 2014, Venesky et al. 2014). However, in some cases,
145 biodiversity–disease outcomes may diverge from each other based on key differences in specific
146 host traits or pathogen characteristics (Becker et al. 2014, Wood et al. 2014a, 2014b, Strauss et
147 al. 2015, 2018). Finally, biodiversity–disease relationships may be idiosyncratic and context-
148 dependent on the specific combinations of host and pathogen traits (Salkeld et al. 2013, Wood et
149 al. 2014a, Strauss et al. 2015). Therefore, by simultaneously studying biodiversity–disease
150 relationships for multiple similar pathogens each infecting multiple related host species, we can
151 look for common patterns among many host–pathogen pairs and identify potential host or
152 pathogen traits that lead to different outcomes.

153 Pollinator communities are a good system to study biodiversity–disease relationships
154 because many pollinator species are infected by several multi-host pathogens that may be
155 affected by community-level factors in different ways. Three related viruses, deformed wing
156 virus (DWV), black queen cell virus (BQCV), and sacbrood virus (SBV), have long been
157 observed in honey bees (*Apis mellifera*). The same viral strains that infect honey bees also spill-
158 over into other native bee species, but initial evidence suggests that native bees are less
159 commonly infected compared to honey bees and may be less competent hosts (Singh et al. 2010,
160 Fürst et al. 2014, Manley et al. 2015, McMahon et al. 2015). Current evidence suggests that the
161 viruses may be transmitted through contact with flowers shared among pollinators, particularly
162 through contaminated pollen (Singh et al. 2010, McArt et al. 2014, Alger et al. 2019). Pollinator
163 species vary substantially in their flower preferences, sociality, and other life history traits
164 (Williams et al. 2010), which could impact the likelihood of pathogen exposure and infection
165 among different hosts and in different community contexts.

166 We measured viral prevalence in pollinator communities to address: 1) How does
167 pathogen prevalence differ among host species and pathogens?, 2) How does pathogen
168 prevalence vary among communities that differ in host species richness, relative abundance, and
169 composition?, and 3) Are relationships between pathogen prevalence and pollinator community-
170 level factors similar among hosts or pathogens? First, we expected that all three viruses would be
171 present in all host species tested, but that managed honey bees, as the main reservoir host, would
172 have higher viral prevalence for all three viruses compared to other native bee species. Second, if
173 pollinator host species vary in virus prevalence, then we predicted that community-level factors,
174 such as pollinator community species richness, abundance, and community composition, would
175 all vary with virus prevalence among different communities. Specifically, we thought that greater
176 species richness would be likely to reduce virus prevalence, while greater pollinator abundance
177 would increase virus prevalence, and communities with similar host compositions would exhibit
178 similar virus prevalence compared to disparate communities. Third, we expected that
179 relationships between virus prevalence and the three community-level factors would show
180 consistent patterns among the three related viruses and four common pollinator hosts.

181 **Methods**

182 Study system

183 Three picorna-like RNA viruses, black queen cell virus (BQCV) in the Dicistroviridae
184 family, and deformed wing virus (DWV) and sacbrood virus (SBV) in the Iflaviridae family,
185 commonly infect European honey bees (*Apis mellifera*) (Chen and Siede 2007). Growing
186 evidence suggests that these viruses are bidirectionally transmitted among managed honey bees
187 and native bees (Singh et al. 2010, Fürst et al. 2014, McMahon et al. 2015, Alger et al. 2019,
188 Grozinger and Flenniken 2019). Though these viruses may be generalist pathogens capable of
189 infecting a wide diversity of species, all three viruses are most commonly found in honey bees
190 and less commonly detected in other native pollinator species (Singh et al. 2010, Manley et al.
191 2015, Dolezal et al. 2016). Viral infections in early life stages (e.g. larval or pupal) cause
192 mortality in honey bees, while infected adults are typically asymptomatic but can still transmit
193 the virus (Chen and Siede 2007, Grozinger and Flenniken 2019). Some native bees may
194 experience reduced viral virulence compared to honey bees (Dolezal et al. 2016), but viral
195 virulence in native bee species has received limited study. Viral transmission among conspecifics
196 is likely food-borne or fecal-oral (Chen and Siede 2007) via contact on flowers (McArt et al.
197 2014). DWV and BQCV have been detected on whole flowers near apiaries and on pollen
198 collected by bees, and honey bees can become infected after consuming virus-contaminated
199 pollen (Singh et al. 2010, Mazzei et al. 2014, Alger et al. 2019).

200 Sampling pollinator communities

201 We collected pollinators from 14 winter squash farms in Michigan, USA, with
202 permission granted by private landowners (Appendix S1: Table S1). All fields were adjacent to
203 either corn or apple orchards, except for the GT and S sites, which had small plots of other
204 specialty vegetables. Field sites were at least 10 km away from each other, so it is unlikely that
205 bees observed at one site visited other field sites. We sampled the pollinator communities at each
206 site twice during the peak squash flower bloom (July and August), and maintained even
207 sampling effort in terms of both total time and area sampled per site. We sampled on sunny days
208 with little cloud cover and wind speeds less than 2 m/s during the peak squash bloom period (18
209 July – 21 August 2015 and 26 July – 2 September 2016).

210 Bees were sampled via hand-netting and pan traps in four 50-m transects. Three transects
211 were randomly placed within the field in line with the crop rows, and one transect was placed
212 along the field edge. Edges typically contained a mixture of native flowers and weeds. We hand-
213 netted pollinators within 1.5-m of each transect line once for 30 minutes at 0800, 1000, 1100 and

214 1200. We did not collect in the afternoon because squash flowers close by midday. Fluorescent
215 blue, yellow, and white pan traps were set along the transect between the crop rows 5-m apart in
216 an alternating color pattern. Pan traps were set prior to 0700 and collected at 1200, after squash
217 flowers close. Pan traps were checked every 3 hours. All insects collected were frozen for later
218 identification and viral analysis. Bee collection method (i.e. netting or pan traps) was not
219 correlated with virus presence or absence (Appendix S1: Table S2).

220 Each specimen was identified using the Discover Life key (<http://www.discoverlife.org>).
221 Most specimens were identified to species. *Lasioglossum* and *Halictus* were identified to genus
222 because they are very difficult to key out to species. Additionally, rare wasp genera with less
223 than five total occurrences in our sample were identified to genus.

224 Detecting viral positive strand prevalence

225 We tested for BQCV, DWV, and SBV within four pollinator species: *Apis mellifera* (n =
226 237), *Bombus impatiens* (n = 252), *Eucera pruinosa* (n = 193), and *Lasioglossum* spp. (n = 255).
227 These four species were the most consistently abundant species among all communities sampled
228 (Appendix S1: Table S3). We tested up to 20 randomly selected individuals from each species
229 per site, and tested all individuals available when less than 20 were collected at a site (Appendix
230 S1: Table S4).

231 Tissue from half of each bee's abdomen was homogenized using a FastPrep-24 (MP
232 Biomedicals, Santa Clara, CA, USA) for 1 minute at 4.0 M/sec. RNA was extracted using
233 TRIzol reagent (Ambion, Austin, TX, USA) according to manufacturer's instructions, eluted in
234 30 μ l DNase/RNase free H₂O, and RNA concentration was quantified using Qubit 3.0
235 Fluorometer (Invitrogen, Carlsbad, CA, USA). We found that RNA concentration did not impact
236 the likelihood of detecting viral presence (Appendix S1: Table S2, Appendix S2: Section S1).
237 Positive strand complimentary DNA (cDNA) synthesis reactions were performed with 2 μ l of
238 RNA template in a 20 μ l reaction using M-MLV reverse-transcriptase (Promega, Madison, WI,
239 USA) and 0.25 μ M random hexamers (Invitrogen) according to manufacturer's instructions.

240 We tested for the presence or absence of BQCV, DWV, and SBV positive strand using
241 PCR with established virus-specific primers (Appendix S1: Table S5). The DWV primer did not
242 differentiate between DWV-A, -B, or -C variants, therefore reported DWV prevalence includes
243 all three variants. All reactions included negative (H₂O) and virus-specific positive controls. To
244 confirm adequate RNA extraction and reverse transcription of all bee samples, we ran PCR for

245 each sample with *A. mellifera* 18S rRNA gene primers (Cardinal et al. 2010) as a control. Further
246 reaction details are provided in Appendix S2: Section S1. All PCR products were visualized with
247 gel electrophoresis to determine virus presence or absence. We sequenced a subset of the PCR
248 products to confirm identification of viral RNA and the 18S gene (GenBank Accession Numbers
249 in Appendix S1: Table S6).

250 The BQCV, DWV, and SBV prevalence observed in this study are representative of
251 current spillover among pollinator species. The primers we used were created from honey bee
252 virus sequences (Singh et al. 2010), so they could slightly underestimate the virus prevalence in
253 native bees. However, data from several studies indicate that native bees share the same virus
254 strains with local honey bees (Singh et al. 2010, Genersch et al. 2011, Yang et al. 2013, Levitt et
255 al. 2013, Fürst et al. 2014, McMahon et al. 2015, Radzevičiūtė et al. 2017, Bailes et al. 2018).
256 Further, the primers we use are well established for successfully testing for viral positive and
257 negative strand presence in many bee, wasp, and non-Hymenopteran insect species (Singh et al.
258 2010, Levitt et al. 2013, Fürst et al. 2014, McMahon et al. 2015, Bailes et al. 2018).

259 Screening for the viral negative strand

260 We determined the infection status of a subset of virus-positive samples with additional
261 negative-strand specific RT-PCR. Identifying the negative strand provides strong evidence of
262 viral replication and an active infection within the host (Ongus et al. 2004, Yue and Genersch
263 2005). Up to 26 virus-positive bee samples from each of the focal bee species per virus were
264 randomly selected from all sites to test for the presence of the negative strand. If fewer than 20
265 virus-positive bee samples for a species were available, then all virus-positive samples were used
266 (Appendix S1: Table S7). Negative-strand specific cDNA synthesis was carried out with 2.5 µl
267 RNA template with M-MLV reverse transcriptase (Promega) and tagged negative-strand specific
268 primers for BQCV, DWV, and SBV, followed by PCR with negative and virus-specific positive
269 controls (primer details in Appendix S1: Table S5). All samples were visualized with gel
270 electrophoresis, and a subset of samples were sequenced to confirm identification of the negative
271 strand viral sequences (GenBank Accession Numbers in Appendix S1: Table S6). Additional
272 reaction details are in Appendix S2: Section S2.

273 Statistical analysis

274 All analyses were performed in the program R v4.0.2 (R Core Team 2020). We used a
275 global Generalized Linear Mixed effects model (GLMM) of virus prevalence including all three
276 viruses within the four host species with a binomial distribution and logit link function (*lme4*
277 package) (Bates et al. 2015). Here, we use ‘virus prevalence’ as the response variable in our
278 global model based on the presence or absence of the viral positive strand for each individual
279 bee. The ‘infection prevalence’, based on the presence of the viral negative strand, had
280 insufficient sample size among hosts and sites to be used in the global model (see Appendix S2:
281 Section S3.1 for further discussion). For random effects, we included visit number nested within
282 site to account for bees collected from sites on different days, and each bee’s unique ID to
283 account for testing each bee for BQCV, DWV, and SBV. All models included species richness,
284 total pollinator abundance, virus type (BQCV, DWV, and SBV), and host species (*A. mellifera*,
285 *B. impatiens*, *Lasioglossum* spp., and *E. pruinosa*) as main effects. Total pollinator abundance
286 was log transformed, and all continuous variables were z standardized. We evaluated the model
287 without interactions (Model 1) and each combination of two- (Models 2a–f), three- (Models 3a–
288 e), and four-way interactions (Model 4) in a model selection table ranked by lowest AICc score
289 (*MuMIn* package; Table 1, Appendix S1: Table S8, top model selection details in Appendix S2:
290 Section S3.2) (Barton 2020). Significant main effects do not differ between any of the top
291 models, indicating that our key results are robust.

292 All top models included a significant interaction between virus type and host species.
293 Interaction effects in non-linear GLMMs are complicated and cannot simply be evaluated by the
294 coefficient or significance of the interaction term (Ai and Norton 2003). Instead, we investigated
295 the asymptotic variance of the interaction using a post-hoc pairwise comparison of predicted
296 virus prevalence among each host species for each virus with a Tukey method for adjusting the
297 p-value for multiple comparisons (package *emmeans*) (Lenth 2020). We also conducted a Type II
298 Wald Chi-square test to construct an Analysis of Deviance table for the main factors in Model 2a
299 and Model 3a (package *car*; Table 3, Appendix S1: Table S9) (Fox and Weisberg 2019). All
300 factors in the top model had Variance Inflation Tests (VIF) < 6, below the standard threshold of
301 10 for collinearity issues (Appendix S1: Table S10) (Dormann et al. 2013). Furthermore, we
302 compared the results from the top Model 2a to a model that included *A. mellifera*, *B. impatiens*,
303 *Lasioglossum* spp., and *E. pruinosa* specific abundances (log transformed) instead of total
304 abundance, and found similar results to Model 2a (Appendix S1: Table S11). Viral prevalence

305 was not associated with any of the four focal host's species-specific abundances. However, we
306 did not have the power to adequately test the effect of the abundance of all potential host species
307 on virus prevalence because rarer species were not consistently found at all sites.

308 There was no evidence of significant spatial autocorrelation in the model residuals for
309 any model, indicating that closely located communities did not have significantly similar virus
310 prevalence (Moran's I test using packages *ape* and *DHARMA*; Appendix S1: Table S12) (Paradis
311 and Schliep 2018, Hartig 2020). Therefore, we considered virus prevalence among different
312 pollinator communities as independent of each other.

313 To calculate apparent 'infection prevalence' (based on the presence of viral negative
314 strand) within each host species, we used the 'epi.prev' function in the *epiR* package (Stevenson
315 et al. 2020). The negative strand infection prevalence is determined by the number of samples
316 with the viral negative strand present divided by the number of virus-positive samples that were
317 tested, which indicates active replication in the host (Ongus et al. 2004, Yue and Genersch
318 2005). We compared negative strand infection prevalence in each of the four host species within
319 each virus using a Chi-squared test of two proportions. We used a Bonferroni correction for
320 multiple comparisons to determine significant differences among host species ($\alpha^* = 0.05/6 =$
321 0.0083). The Chi-squared test approach achieved similar results when compared with the GLMM
322 post-hoc analysis comparing differences in virus prevalence (positive strand) among the four
323 host species (described above).

324 Species richness, Simpson's diversity index (1-D), and species-specific and total
325 abundance for each pollinator community were determined from the collection data for each site.
326 Community composition was assessed qualitatively through differences in the relative abundance
327 of pollinator species and Non-metric Multi-Dimensional Scaling (NMDS) (described below). We
328 tested the nested temperature of the pollinator communities sampled compared to simulated null
329 model communities following Johnson et al. 2013b (method: "r00", function 'oecosimu',
330 package *vegan*; Appendix S1: Fig. S1) (Oksanen et al. 2018). To determine if we captured the
331 pollinator species richness within each community, we created individual-based rarefaction
332 curves (*iNext* package) and compared the observed species richness to the estimated species
333 richness at the asymptote of the rarefaction curve (Appendix S1: Fig. S2) (Hsieh et al. 2016). For
334 invertebrate communities, it is rare that the observed species richness ever reaches an asymptote
335 (Novotný and Basset 2000, Gotelli and Colwell 2001). Although observed and estimated species

336 richness differed, there was strong consistency in the ranking order of the communities
337 (Appendix S1: Table S13). Additionally, we found that our results were robust regardless of
338 method used to estimate species richness because models with two different methods of
339 estimating species richness showed the same results as Model 2a (Appendix S1: Table S14 and
340 S15, details in Appendix S2: Section S3.3). Therefore, the observed species richness seemed to
341 sufficiently describe differences among the pollinator communities based on our even sampling
342 effort in both time spent sampling and area covered by transects at each site.

343 To examine how community composition influenced virus prevalence in different host
344 species, we used Non-metric Multidimensional Scaling (NMDS) ordination of all pollinator
345 species identities and relative abundances collected at each site. Specifically, this analysis
346 examines whether other community members beyond the four focal host species may be
347 indicator species correlated with higher virus prevalence by evaluating the presence/absence and
348 relative abundance of all pollinator species in the community. We predict that communities that
349 include a key indicator species will show a consistent correlation with high virus prevalence, but
350 we expect that rare and low-density pollinator species are unlikely to show significant correlation
351 with virus prevalence. The NMDS ordination of the pollinator communities was created using a
352 Bray-Curtis dissimilarity matrix (*vegan* package) (Oksanen et al. 2018). A two-dimensional
353 solution for the NMDS ordination of pollinator community composition yielded a stress value of
354 0.1324, which showed that the 2D fit corresponded well with the actual multivariate distance
355 among communities and was well below the 0.2 stress threshold.

356 We separately evaluated the correlation between BQCV, DWV, and SBV prevalence
357 within each of the four host species and the ordination of pollinator communities using fitted
358 smooth surfaces (i.e. contour lines) calculated using Generalized Additive Models (GAM) with
359 thin-plate splines ('ordisurf' function; *vegan* package). The correlation between host–virus
360 prevalence and pollinator community composition were evaluated with GAM fitted vectors that
361 indicate the strongest linear gradient along the fitted contour lines of virus prevalence in the
362 ordination (adjusted R^2). By comparing patterns of virus prevalence and directionality of the
363 fitted vectors overlaid on the NMDS plots of pollinator community composition for each host–
364 virus pair, we can determine whether communities with similar compositions tend to share
365 patterns of virus prevalence (additional details in Appendix S2: Section S3.4).

366 **Results**

367 1) How does pathogen prevalence differ among host species and pathogens?

368 *Virus and infection prevalence were highly variable among honey bees and native bees*

369 The BQCV, DWV, and SBV positive strands were detected in the four focal pollinator
370 species: *Apis mellifera*, *Bombus impatiens*, *Lasioglossum* spp., and *Eucera pruinosa* (Fig. 1,
371 Appendix S1: Table S16 and S17). Furthermore, virus prevalence varied significantly among the
372 three viruses and different host species, as all the top generalized linear mixed models (GLMM)
373 from model selection included a significant interaction between virus type and host species (Fig.
374 1, Table 1, Appendix S1: Table S18). BQCV and DWV had the same overall pattern of
375 prevalence among the four host species tested, where *A. mellifera* had significantly higher
376 prevalence than *B. impatiens*, which in turn was significantly higher than both *Lasioglossum* spp.
377 and *E. pruinosa* (Fig. 1). SBV prevalence showed a different pattern among the four host
378 species. *A. mellifera* and *B. impatiens* had similar SBV prevalence, but SBV was extremely rare
379 in *Lasioglossum* spp. and *E. pruinosa* (estimated 0.2% and 1.1% prevalence by Model 2a,
380 respectively).

381 We also tested the prevalence of viral infection by testing for BQCV, DWV, and SBV
382 negative strand in each host species (hereafter, 'infection prevalence'). The viral negative strand
383 for all three viruses was present in all four host species, except for SBV in *Lasioglossum* spp.
384 (Table 2). *Lasioglossum* spp. had very low SBV prevalence detected (0.2%, a single SBV-
385 positive individual), so it is unsurprising that we found no evidence of the SBV negative strand.

386 The patterns of infection prevalence varied among the pollinator hosts and viruses. In
387 general, virus-positive *A. mellifera* and *B. impatiens* had higher infection prevalence compared to
388 *Lasioglossum* spp. and *E. pruinosa* (Table 2, Appendix S1: Table S19). The infection prevalence
389 presented here was an estimate since we only tested a subset of virus-positive specimen from
390 each species, but the data clearly showed that there was variation in the likelihood of infection
391 among host species for all three viruses.

392 2) How does pathogen prevalence vary among communities that differ in host species richness,
393 relative abundance, and composition?

394 *Pollinator communities varied in abundance, richness, and composition*

395 Across both sampling years, we collected 4,737 bees and wasps from 14 communities,
396 including at least 126 species and 78 genera from five bee families (Andrenidae, Apidae,

397 Colletidae, Halictidae, and Megachilidae) and nine wasp families (Aulacidae, Crabonidae,
398 Gasteruptionidae, Ichneumonidae, Pompilidae, Sphecidae, Thynnidae, Tiphiidae, and Vespidae).
399 The most common genera were *Lasioglossum* (n = 1305), *Bombus* (n = 1071), *Eucera* (n = 843),
400 *Apis* (n = 508), *Vespula* (n = 129), *Augochlora* (n = 127), and *Halictus* (n = 105). The pollinator
401 communities varied in species richness (range: 7 to 49 species) and total pollinator abundance
402 (range: 46 to 756 individuals) (Fig. 2). Furthermore, pollinator community composition varied
403 qualitatively among sites, as the relative abundance of key pollinator species differed among
404 communities (Fig. 2, Appendix S1: Fig. S3). The pollinator communities were significantly
405 nested compared to simulated null community matrices, such that species poor communities
406 were composed of a subset of the species rich communities (observed nested temperature =
407 20.7°; average null model temperature = 54.2°, p = 0.01; Appendix S1: Fig. S1). All
408 communities included *A. mellifera*, *B. impatiens*, *Lasioglossum* spp., and *E. pruinosa*, except *E.*
409 *pruinosa* was absent from K site. Simpson's index of diversity (1-D) ranged from 0.46 to 0.85
410 among the different communities (Appendix S1: Table S13).

411 *Virus prevalence was linked with pollinator species richness, but not pollinator abundance nor*
412 *community composition*

413 Virus prevalence was more strongly associated with pollinator species richness than with
414 other community characteristics, like total host abundance or species-specific abundances.
415 Pollinator community species richness was a significant main effect in the top GLMM (Model
416 2a; Table 3). Specifically, all four host species had significantly reduced DWV prevalence in
417 communities with greater pollinator species richness (Fig. 3a). Additionally, *A. mellifera* and *B.*
418 *impatiens* had significantly reduced BQCV and SBV prevalence in species-rich communities
419 (Fig. 3a). *Lasioglossum* spp. and *E. pruinosa* had relatively low BQCV and SBV prevalence
420 among all communities tested, and therefore did not show as much variation in viral prevalence.
421 On the other hand, total pollinator abundance and the species-specific abundances of *A.*
422 *mellifera*, *B. impatiens*, *Lasioglossum* spp. and *E. pruinosa* were not significant predictors of
423 virus prevalence in any of the top models (Fig. 3b, Table 3, Appendix S1: Table S11).

424 Pollinator community composition generally did not predict virus prevalence in most host
425 species. The NMDS ordination was only significantly correlated with viral prevalence in two of
426 the twelve host–virus pairs, specifically *A. mellifera* SBV prevalence and *Lasioglossum* spp.

427 DWV prevalence (*A. mellifera* SBV: $F = 3.02$, $p = 0.03$, $\text{Adj } R^2 = 0.68$; *Lasioglossum* spp.
428 DWV: $F = 2.15$, $p = 0.02$, $\text{Adj } R^2 = 0.60$; Appendix S1: Fig. S4a and S4c).

429 3) Are relationships between pathogen prevalence and community-level factors similar among
430 hosts or pathogens?

431 *Consistent relationships between virus prevalence and pollinator community species richness*
432 *and abundance in hosts and pathogens*

433 All three viruses showed significantly reduced virus prevalence in species-rich
434 communities within host species that had greater than 10% estimated virus prevalence (Fig. 3a).
435 The strength of the negative relationships varied among host species based on their relative viral
436 prevalence. BQCV and SBV showed clear negative slopes between virus prevalence and species
437 richness in *A. mellifera* and *B. impatiens*, hosts with high BQCV and SBV prevalence.
438 Meanwhile, *Lasioglossum* spp. and *E. pruinosa* were rarely infected with BQCV or SBV, and
439 showed no strong relationship between virus prevalence and species richness (Fig. 1, Fig. 3a).
440 None of the host–virus pairs had greater virus prevalence in species-rich communities.

441 When comparing across either hosts or viruses, virus prevalence was largely unlinked
442 with community composition. In two of the twelve host–virus pairs, there were significant
443 relationships between viral prevalence and community composition, but the direction of the
444 relationships varied (Appendix S1: Fig. S4).

445 There were no significant relationships between virus prevalence and pollinator
446 community total abundance among all host species and viruses tested (Table 3, Fig. 3b,
447 Appendix S1: Table S11).

448 **Discussion**

449 Species richness is the most important community factor associated with reduced
450 pathogen prevalence across multiple hosts and multiple pathogens. In contrast, host abundance
451 and community composition are not consistently associated with pathogen prevalence. This work
452 illustrates the dilution effect pattern for pollinator viruses for the first time. For multiple viruses
453 within multiple bee host species, communities with greater pollinator species richness had lower
454 viral prevalence than species-poor communities, but the strength of the relationships appear to
455 vary based on each species' competence for each virus.

456 Species richness

457 Increasingly biodiversity–disease studies have begun to focus on multi-host–pathogen
458 systems to evaluate how disease risk within different host species respond to changes in host
459 communities. However, investigations that simultaneously compare biodiversity–disease
460 relationships in multiple pathogens that infect similar communities of hosts have been much
461 rarer (but see Johnson et al. 2013a). Here, we find that pollinator communities with greater
462 species richness exhibit consistently lower virus prevalence for three multi-host viruses within
463 four focal bee species, while controlling for total host abundance (Fig. 3). Broadly, our findings
464 corroborate other multi-host pathogen studies that have found consistent patterns of dilution in
465 pathogen prevalence among multiple co-occurring hosts or community-wide pathogen
466 prevalence (Ezenwa et al. 2006, Johnson et al. 2013a, 2013b, Becker et al. 2014, Venesky et al.
467 2014, Strauss et al. 2018).

468 The pollinator–virus system has many characteristics that typically facilitate the dilution
469 effect in other host–pathogen systems. The dilution effect is likely to occur when the most
470 competent host dominates species-poor communities, and more disease resistant host species are
471 common in species-rich communities (LoGiudice et al. 2003, Keesing et al. 2006, Johnson et al.
472 2013b). Biodiversity is lost from pollinator communities in a non-random order, where solitary
473 and specialist native bees tend to be extirpated first (Rader et al. 2014). Our results are consistent
474 with this pattern, as pollinator communities in our study are nested (Appendix S1: Fig. S1).
475 Species poor communities are dominated by the four focal hosts in our study, two of which (*A.*
476 *mellifera* and *B. impatiens*) are competent hosts with high prevalence for all three viruses (Fig.
477 1). Species-rich communities include many native bee species, which are likely to be less or non-
478 competent viral hosts (Singh et al. 2010, Manley et al. 2015, Dolezal et al. 2016).

479 Our results suggest that the encounter reduction mechanism of the dilution effect may
480 operate in the pollinator–virus system. Specifically, species-rich host communities may have
481 lower encounter rates between susceptible hosts and infectious viral particles or infected hosts
482 due to a higher proportion of non-hosts or low competence hosts in species-rich communities
483 (Keesing et al. 2006). As highly competent hosts and floral generalists, *A. mellifera* and *B.*
484 *impatiens* may disproportionately impact virus prevalence in species-poor communities by
485 spreading viral particles to more flowers and increasing the likelihood of hosts encountering viral
486 particles during visits to shared flowers (i.e. encounter reduction) (Keesing et al. 2006). Also, if

487 native bee hosts in species-rich communities can act as decoys or “diluter hosts” that take up
488 viral particles but do not become infected during visits to shared flowers, then susceptible hosts
489 could have a reduced encounter rate with viral particles (Johnson and Thielges 2010). Further
490 investigation through paired experimental and natural studies is needed to elucidate the specific
491 dilution effect mechanism(s) operating in pollinator pathogen systems and to improve future
492 predictions of disease risk.

493 Species abundance

494 Community factors other than biodiversity were not strongly associated with virus
495 prevalence, including total pollinator abundance and the species-specific abundances of the four
496 focal host species (Fig. 3b, Table 3, Appendix S1: Table S11). Changes in community diversity
497 often correspond with changes in the total host abundance and/or relative abundance of specific
498 host species, which can lead to the ‘susceptible host regulation’ mechanism of the dilution effect
499 (Rudolf and Antonovics 2005, Randolph and Dobson 2012, Mihaljevic et al. 2014). Susceptible
500 host regulation could operate in the pollinator–virus system if additional low competence hosts
501 compete with susceptible hosts to constrain their abundance and reduce pathogen spread
502 (Keesing et al. 2006). Most of the other pollinator species in these communities were rare (less
503 than 5 individuals observed per site) and are unlikely to explain community-level differences in
504 virus prevalence (see Appendix S1: Fig. S4 for an analysis that considers additional pollinator
505 species). Further, we found no relationship between pollinator host abundance and virus
506 prevalence over all, so susceptible host regulation is unlikely to mediate the dilution effect.

507 The lack of relationship between host abundance and viral prevalence suggests that
508 BQCV, DWV, and SBV may have frequency-dependent transmission rather than density-
509 dependent transmission. The three viruses are likely transmitted within and among host species
510 through interactions on flowers and through contaminated pollen (Singh et al. 2010, McArt et al.
511 2014, McMahon et al. 2015). As a result, viral transmission may depend on the frequency of
512 pollinator visits to shared flowers rather than the abundance of pollinators in a community.
513 Pathogens with frequency-dependent transmission are more likely to exhibit decreased pathogen
514 prevalence with greater community biodiversity (i.e. dilution effect) that is not influenced by the
515 total number of hosts in the community (Rudolf and Antonovics 2005, Keesing et al. 2006).
516 Future studies should explicitly examine the mode of transmission of pollinator viruses and

517 whether the frequency of bee contacts with flowers provide a better fit with patterns of pathogen
518 prevalence among different pollinator communities than host abundance.

519 Community composition

520 Pollinator community composition was rarely found to influence virus prevalence among
521 most host–virus pairs tested. This is interesting because community composition is an important
522 driver of observed dilution effects in many host–pathogen systems (Roche et al. 2012, Johnson et
523 al. 2013b, Salkeld et al. 2013, Becker et al. 2014). Assuming that hosts species are not equally
524 competent for a pathogen, the presence or absence of a particular species in a community can
525 dramatically influence pathogen transmission dynamics. This process could be akin to the
526 “selection effect” from the field of biodiversity–ecosystem function (BEF), where a particular
527 species has a disproportionate impact on pathogen prevalence and/or transmission in species-rich
528 communities, which could lead to either dilution or amplification effects depending on the host
529 species’ traits (Loreau M. and Hector. A. 2001). However, virus prevalence among all four
530 pollinator species was generally unrelated to community composition.

531 Community composition may also influence virus prevalence if the presence of particular
532 pollinator species influences the likelihood of viral encounter or transmission by altering
533 interactions among host species on shared flowers. Though our study does not evaluate the
534 “complementarity effect” mechanism from BEF literature, where pathogen transmission is
535 reduced through less habitat sharing among host species in diverse communities, it could occur
536 in pollinator pathogen systems (Loreau M. and Hector. A. 2001, Becker et al. 2014). Bees in
537 diverse communities may reduce their shared flower use through greater specialization in
538 foraging or utilize different parts of the flower (e.g. nectar vs. pollen), which could reduce the
539 potential for viral encounter or transmission among species through a complementarity
540 mechanism. Future work needs to investigate how specific pollinator interactions on flowers
541 among different communities contribute to various dilution effect mechanisms.

542 Consistent evidence of dilution among pathogens and hosts

543 We found similar, negative biodiversity–disease relationships among multiple viruses and
544 multiple hosts, but the strength of the dilution effect varied among hosts. Variation in the
545 strength of relationships between biodiversity and pathogen prevalence is likely due to variation
546 in relative viral competence among different host species. *A. mellifera* and *B. impatiens*, the two

547 most highly competent hosts in our study displayed consistent dilution effects for all three
548 viruses. Meanwhile *Lasioglossum* spp. and *E. pruinosa* are relatively less competent hosts for
549 DWV, and have a weaker dilution effect compared to *A. mellifera* and *B. impatiens*. For BQCV
550 and SBV, *Lasioglossum* spp. and *E. pruinosa* are poor hosts with extremely low virus
551 prevalence, and consequently there was little virus prevalence to dilute with greater community
552 biodiversity. The four host species differ in their social behavior and whether they are floral
553 specialists or generalists. Both factors may influence variation viral exposure and prevalence,
554 and result in variable strength in the observed dilution effects among hosts.

555 Perhaps we found similar biodiversity–disease relationships among pathogens because
556 the three viruses are quite similar. The three viruses are closely related (order Picornavirales,
557 DWV and SBV from *Iflavirus* genus), predominantly infect Hymenopteran insects (bees and
558 wasps), particularly honey bees (*A. mellifera*), and have similar modes of infection (i.e. fecal-oral
559 and food-borne) (Chen and Siede 2007, McMahon et al. 2018). Similarly, Johnson et al. also
560 found consistently reduced infection success with greater host diversity for five out of seven
561 trematode parasites that share many pathogen characteristics (Johnson et al. 2013a). Previous
562 studies and meta-analyses have compared biodiversity–disease relationships among highly
563 divergent pathogens, generally finding that pathogen ecology, transmission mode, infectivity, or
564 degree of host specialization influence these relationships (Randolph and Dobson 2012, Salkeld
565 et al. 2013, Wood et al. 2014a, 2014b, Rohr et al. 2020). Utilizing a comparative approach for
566 multiple pathogens within ‘replicate’ host communities will clarify how differences in either host
567 or pathogen ecology may dictate variation in biodiversity–disease relationships.

568 Virus prevalence in pollinators

569 Our virus prevalence results are consistent with other studies that found BQCV, DWV,
570 and SBV are shared among many pollinator species (Singh et al. 2010, Fürst et al. 2014,
571 McMahon et al. 2015, Dolezal et al. 2016). However, our study design more accurately assesses
572 differences in BQCV, DWV, and SBV prevalence by using larger sample sizes per species. The
573 results show that *A. mellifera* are highly susceptible and competent hosts for all three viruses. *B.*
574 *impatiens*, a close relative of *A. mellifera*, was also a relatively competent host for all three
575 viruses, but had lower DWV and BQCV prevalence. The more distantly related *E. pruinosa* and
576 *Lasioglossum* spp. have lower viral and infection prevalence, suggesting that both are likely poor
577 hosts, less susceptible, and/or less likely to encounter infective viruses.

578 BQCV, DWV, and SBV appear to vary in their host ranges from generalist to relatively
579 specialist pathogens that primarily infect very closely related hosts. DWV appears to be the
580 broadest generalist pathogen of the three, causing active infections in a wide range of
581 Hymenoptera (Singh et al. 2010, Manley et al. 2015, Dolezal et al. 2016). Meanwhile, SBV has
582 the most restrictive host range limited primarily to honey bees and bumblebees (*Bombus* spp.),
583 and BQCV is intermediate between the two (Manley et al. 2015). Despite some differences in
584 host range, all three viruses showed very similar biodiversity–disease relationships.

585 Limitations and future directions

586 Although our findings show intriguing patterns among pollinator communities and
587 pathogen prevalence, they are inevitably limited in scale. Communities are rarely static through
588 time and space as host species vary in phenology, behavior, home ranges, and migration patterns,
589 which consequently can alter expected outcomes for biodiversity–disease relationships (Estrada-
590 Peña et al. 2014, Rohr et al. 2020). In particular, pollinator species vary in their phenology from
591 short (less than a month) to long (the full growing season) (Burkle et al. 2013), and in their
592 specific foraging and nesting habitat requirements (Williams et al. 2010), which result in highly
593 dynamic pollinator communities through time and space. Repeated temporal sampling of a few
594 sites showed that pollinator community diversity declined throughout the growing season, but
595 *Nosema* spp. and *Crithidia* spp. parasite prevalence increased with greater *A. mellifera* and
596 *Bombus* spp. dominance in the communities (Graystock et al. 2020). Our study provides an
597 initial investigation of biodiversity–disease relationships for pollinator viruses toward the end of
598 the growing season and across many similar sites with variable surroundings. Future studies that
599 examine these relationships over different spatial scales and with repeated temporal sampling
600 will be critical for understanding the context-dependence of biodiversity–disease relationships in
601 pollinator–pathogens (Johnson et al. 2015, Graystock et al. 2020, Rohr et al. 2020).

602 Conclusions

603 Overall, prevalence of three viruses in pollinator communities was most strongly linked
604 with species richness, while host abundance and community composition were rarely associated
605 with virus prevalence. Notably, virus prevalence was consistently negatively associated with
606 greater species richness—providing evidence of the dilution effect in multiple viruses infecting
607 multiple pollinator host species. However, we found that the strength of the biodiversity–disease

608 relationships varied based on relative viral prevalence in each host. Host species with high virus
609 prevalence exhibited dilution effects, while hosts with very low virus prevalence did not show a
610 clear biodiversity–disease relationship. Few empirical studies have compared biodiversity–
611 disease relationships among multiple pathogens infecting multiple hosts. We show that this is a
612 powerful approach to assess commonalities and differences in biodiversity–disease relationships
613 within natural systems. Incorporating more realistic complexity of multi-host–multi-pathogen
614 systems into community–disease ecology will improve our understanding of underlying
615 mechanisms that drive differences in pathogen prevalence.

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629

630 **Supporting Information**

631 Additional supporting information may be found online at: [link to be added in
632 production]

633

634 **Data Availability**

635 All data and code used for the analyses and figures contained in this manuscript are
636 available on Dryad (Fearson and Tibbetts 2021): <https://doi.org/10.5061/dryad.zpc866t7g>.

637

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827

828 **Tables**

829 **Table 1:** Model selection table comparing top four models based on lowest AICc. The simpler
 830 Model 2a was selected (bolded) as the top model based on very close performance compared
 831 with Model 3a, but with only a single interaction term rather than a three-way interaction and
 832 three two-way interactions. The full model selection table can be found in Appendix S1: Table
 833 S8, and model results for Model 2a and Model 3a in Table 3 and Appendix S1: Table S9,
 834 respectively.

Model	Model details	K	logLik	AICc	delta	weight
Model 3a	Abundance + Richness × Virus Type × Host Species	27	-1206.03	2466.61	0.00	0.420
Model 2a	Abundance + Richness + Virus Type × Host Species	16	-1217.45	2467.10	0.49	0.328
Model 3c	Abundance × Richness + Virus Type × Host Species	17	-1217.04	2468.31	1.69	0.180
Model 3b	Richness + Abundance × Virus Type × Host Species	27	-1207.79	2470.13	3.52	0.072

835
 836 **Table 2:** DWV, BQCV, and SBV infection prevalence for *Apis mellifera*, *Bombus impatiens*,
 837 *Lasioglossum* spp., and *Eucera pruinosa* determined by the percent of virus-positive samples that
 838 had the viral negative strand present, indicating active viral infections. The 95% confidence
 839 intervals are in parentheses and data include samples randomly selected from all sites. Specific
 840 sample sizes for each host–virus pair are in Appendix S1: Table S7, and p-values for differences
 841 in infection prevalence are in Appendix S1: Table S19.

Species	DWV	BQCV	SBV
<i>Apis mellifera</i>	26.9% (12.3, 46.5)	87.0% (68.0, 96.4)	96.0% (81.0, 99.8)

<i>Bombus impatiens</i>	68.2% (45.2, 85.5)	66.7% (44.9, 84.8)	88.0% (69.7, 96.7)
<i>Lasioglossum</i> spp.	15.0% (4.2, 36.9)	40.0% (18.6, 66.8)	0.0% (0.0, 95.0)
<i>Eucera pruinosa</i>	10.0% (1.8, 31.6)	7.7% (0.4, 33.7)	50.0% (9.8, 90.2)

842

843 **Table 3:** Analysis of deviance table for the top Model 2a generalized linear mixed effects model
844 (GLMM) output based on the Type II Wald Chi squared test. Factors with significant p-values
845 are bolded.

Main Factors	χ^2	df	P value
Total Abundance	1.71	1	0.1907
Species Richness	12.79	1	0.0003
Virus Type	34.63	2	< 0.0001
Host Species	165.25	3	< 0.0001
Virus Type \times Genus	131.18	6	< 0.0001

846

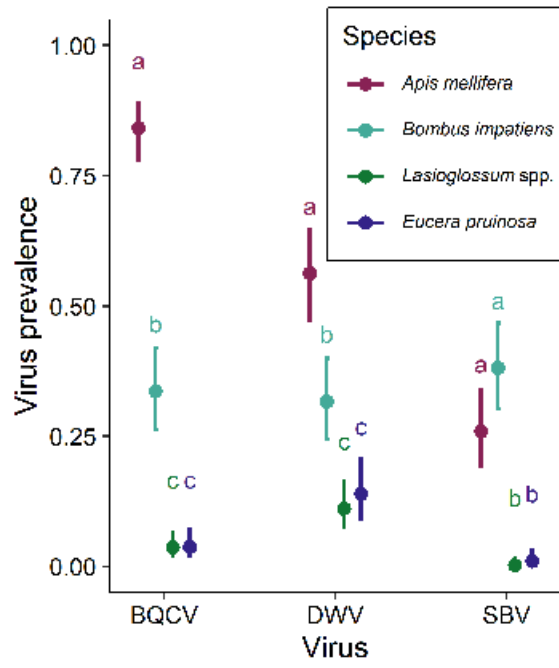
847 **Figure legends**

848 **Fig. 1:** Virus prevalence varied significantly among different host species. BQCV, DWV, and
849 SBV prevalence with the 95% CI among *Apis mellifera*, *Bombus impatiens*, *Lasioglossum* spp.,
850 and *Eucera pruinosa* (Appendix S1: Table S17). Different letters indicate significant differences
851 in virus prevalence among host species and within each virus type. The data shown correspond to
852 the significant virus type \times genus interaction ($p < 0.0001$) from the Model 2a analysis (Table 3),
853 and post-hoc pairwise comparison with a Tukey p-value adjustment for multiple comparisons
854 (Appendix S1: Table S18). Sample sizes per host species: *A. mellifera*, $n = 237$; *B. impatiens*, $n =$
855 252 ; *Lasioglossum* spp., $n = 255$; and *E. pruinosa*, $n = 193$ (Appendix S1: Table S16).

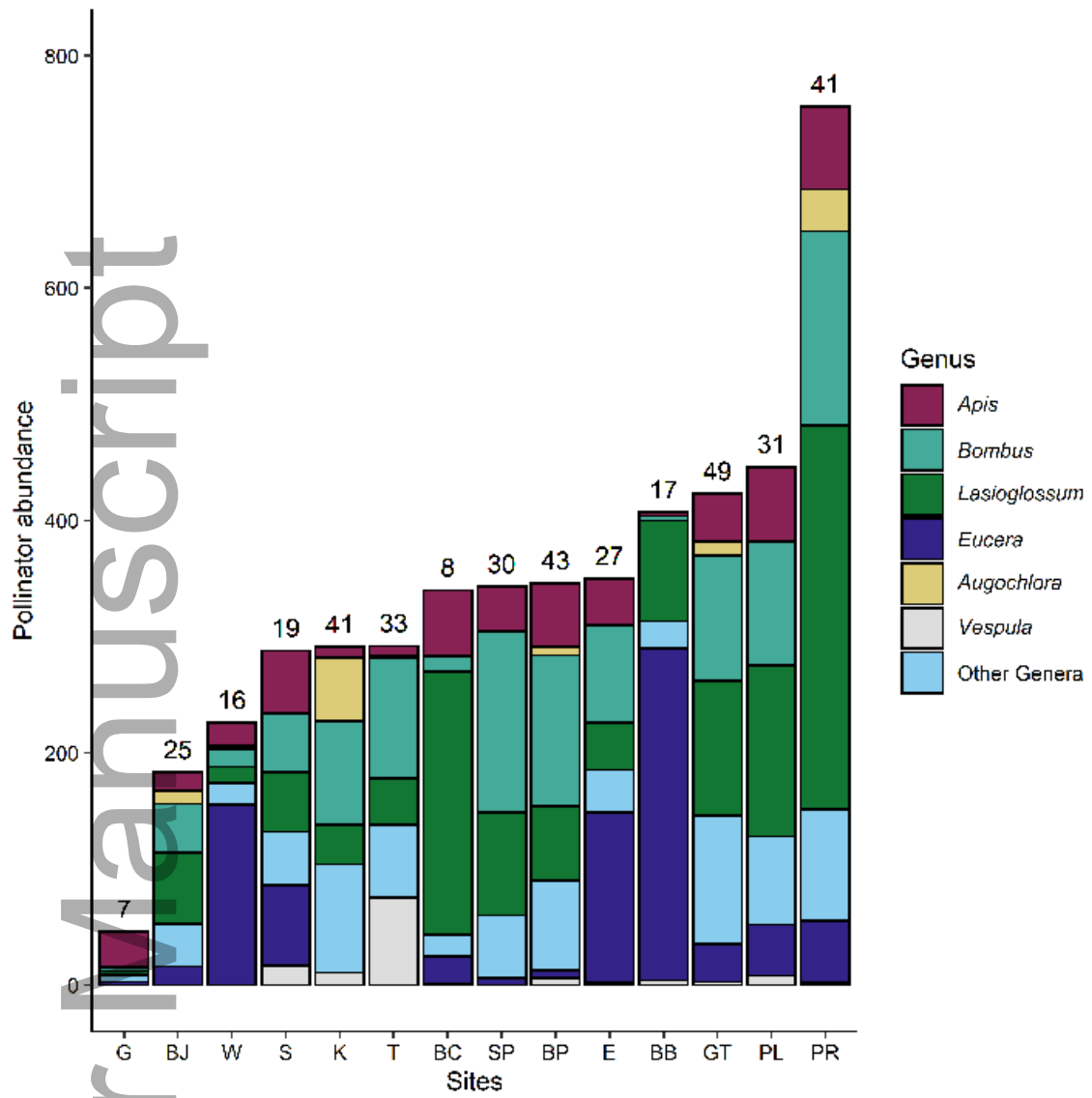
856 **Fig. 2:** Pollinator species richness, abundance, and community composition vary qualitatively
857 among sites. Each bar depicts the relative abundance of the six most common genera and all
858 other genera grouped together per site, with the total height of the bar representing the total
859 pollinator abundance. The observed species richness at each site is shown at the top of each bar.
860 Site abbreviation codes can be found in Appendix S1: Table S1.

861 **Fig. 3:** a) Species-rich communities are significantly correlated with lower predicted virus
862 prevalence in *Apis mellifera*, *Bombus impatiens*, *Lasioglossum*, and *Eucera pruinosa* ($p =$
863 0.0003). The strength of the negative slope varies among host–virus pairs depending on the
864 host’s relative virus prevalence (Fig. 1). b) Total pollinator abundance was not significantly
865 correlated with pollinator virus prevalence ($p = 0.19$). Total pollinator abundance is on a log
866 scale. The data shown correspond to Table 3.

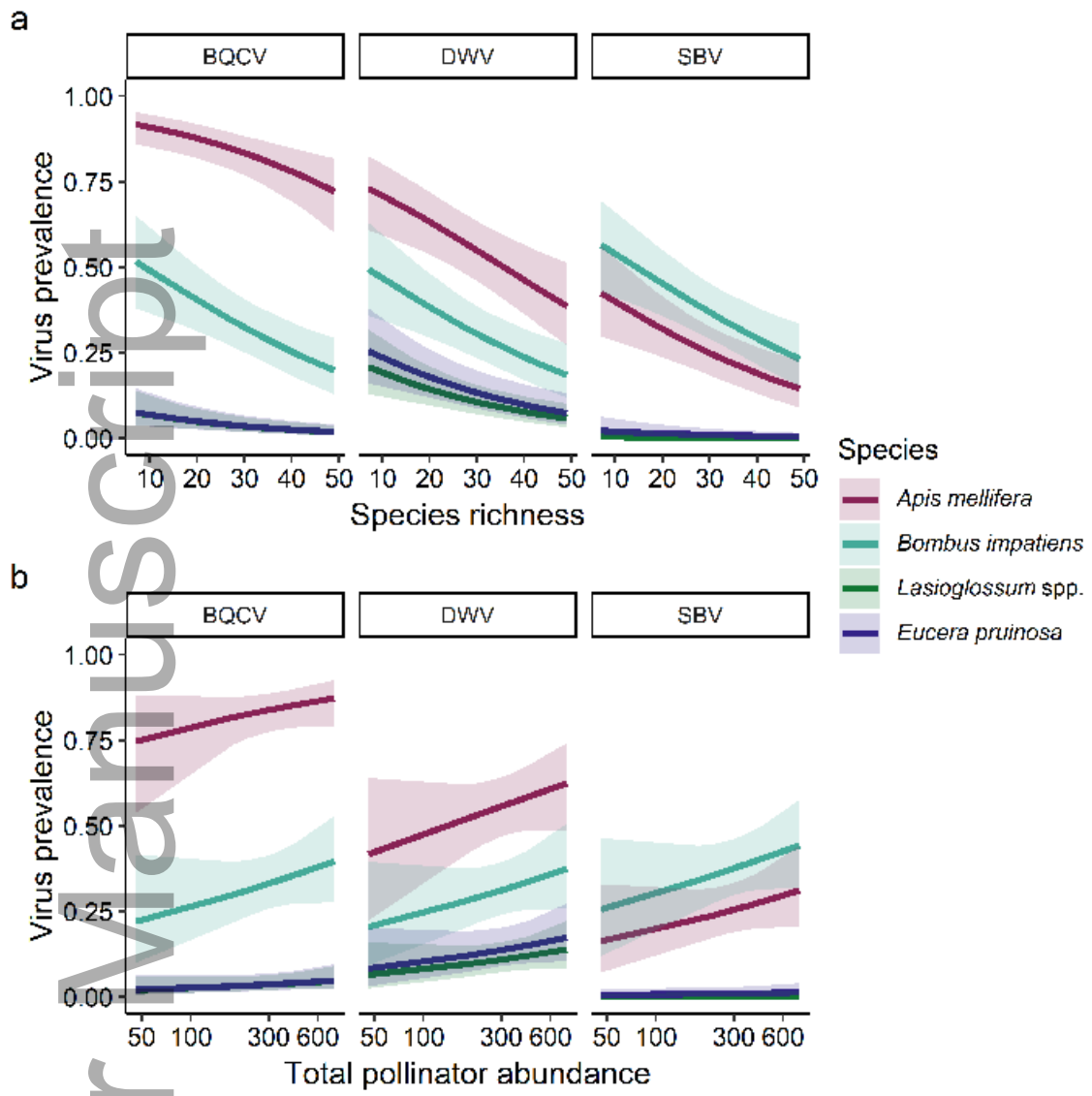
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