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# Intra-guild predation (IGP) can increase or decrease prey density depending on the strength of IGP

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# Abstract

In consumer communities, intra-guild predation (IGP) is a commonly observed interaction that is widely believed to increase resource density. However, some recent theoretical work predicts that resource density should first decrease, and then increase as the strength of IGP increases. This occurs because weak to intermediate IGP increases the IG predator density more than it reduces the IG prey density, so that weak to intermediate IGP leads to the lowest resource density compared to weak or strong IGP. We test this prediction that basal resource density would first decrease and then increase as the strength of IGP increase. We used a well-studied system with two protozoa species engaged in IGP and three bacteria species as the basal resources. We experimentally manipulated the percentage of the IG prey population that was available to an IG predator as a proxy for IGP strength. We found that bacterial density first decreased (by ca. 25%) and then increased (by ca. 30%) as the strength of IGP increased. Using a modified version of a published IGP model, we were able to explain ~70% of the variation in protozoa and bacterial density. Agreement of the empirical results with model predictions suggests that IGP first increased the IG predator density by consuming a small proportion of the IG prey population, which in turn increased the summed consumer density and decreased the bacterial resource density. As IGP strength increased further, the IG predator became satiated by the IG prey, which then freed the bacterial resource from predation and thus increased bacterial density. Consequently, our work shows that IGP can indeed decrease or increase basal resource density depending on its strength. Consequently, the impacts of IGP on resource density is potentially more complex than previously thought.

Keywords: Blepharisma; Colpidium; competition; intra-guild predation; microcosms; population dynamics; predation; protozoa

## 26 Introduction

27 Resource partitioning in space or time has been proposed to be the primary mechanism that  
28 allows consumers to minimize competition for resources (Hutchinson, 1957, 1961, MacArthur,  
29 1958, Schluter, 1993). When consumer species partitioning their resource use, a consumer  
30 community tends to be more efficient in capturing resources and, in turn, reduces resource  
31 density to a lower level (Duffy and Harvilicz, 2001, Finke and Snyder, 2008). However,  
32 there are several types of inter-specific interaction that can either enhance or counter act the  
33 positive effects of resource partitioning on resource capture (Sih et al., 1998). These com-  
34 plex interactions include predator-prey interaction modifications (Sih et al., 1998), predator-  
35 predator facilitation (Losey and Denno, 1998, 1999) and intra-guild predation (IGP; Polis  
36 et al. 1989, Polis and Holt 1992). These more complex interactions also influence how ef-  
37 ficiently prey resources are captured and consumed by a consumer community (Sih et al.,  
38 1998), and need to be considered along with the effects of resource partitioning if we are to  
39 better understand what controls the consumption of prey resources.

40 Among the various types of complex interactions that characterize consumers, intra-  
41 guild predation (IGP) is one of the most widespread and important (Barnes et al., 2018). IGP  
42 occurs when one consumer species (the intra-guild, IG predator) feeds on another one (the  
43 intra-guild, IG prey) with which it also competes for shared basal resources (Polis et al., 1989,  
44 Polis and Holt, 1992). It has been reported that more than half of consumer taxa engage  
45 in IGP across terrestrial and aquatic systems (Arim and Marquet, 2004, Thompson et al.,  
46 2007), and that 50% or more of taxa engage in IGP in the majority of natural communities  
47 (Dunne et al., 2004). The prevalence and uniqueness of IGP in consumer communities make  
48 IGP important for understanding how consumer species and their interactions determine  
49 basal resource consumption.

50 When IGP occurs, most theoretical studies predict that resource density will increase  
51 because consumer assemblages will become less efficient in consuming their basal resources.

52 The majority of these theoretical studies are based on the classic IGP model developed by  
53 Holt and Polis 1997. The classic IGP model predicted that when IGP occurs, the consumer  
54 community would be less efficient in consuming basal resources because the IG prey, which  
55 should be the more efficient consumer, had a lower density due to both competition and  
56 consumption pressure (Holt and Polis, 1997). The prediction that IGP will always increase  
57 resource density remains qualitatively the same in more complex models that include (i)  
58 nonlinear functional responses (Kuijper et al., 2003), (ii) additional species other than IG  
59 prey, IG predator and basal resource (Hart, 2002), (iii) additional trophic supplement to IG  
60 prey or predator (Daugherty et al., 2007), or that (iv) allow IG prey or predator to also prey  
61 on themselves (Rudolf, 2007).

62 But empirical studies have not always born out the expected positive impacts of IGP  
63 on basal resource density. For example, in a meta-analysis, Rosenheim and Harmon showed  
64 that IGP had non-significant effects on basal resource density because more than half (17  
65 out of 29) of the studies showed lower, while the others showed higher, basal resource density  
66 when IGP occurs (Rosenheim and Harmon, 2006). A subsequent meta-analysis found that  
67 basal resource density generally increases with IGP (Vance-Chalcraft et al., 2007); yet, 48%  
68 of studies reviewed also showed decreased density of basal resource when IGP occurs.

69 Why is it that empirical studies have proven heterogeneous, with some finding that IGP  
70 increases prey density, while others find that IGP decreases prey density? Several factors  
71 have been proposed to explain the heterogeneous impacts of IGP on basal resource density.  
72 For example, in ecosystems where exploitative competition is more important than IGP in  
73 governing the population dynamics of IG prey and predator, occurrence of IGP decreases  
74 basal resource density (Vance-Chalcraft et al., 2007). In addition to ecosystem types, some  
75 IG predator species' feeding behavior might induce trait-mediated effects on IG prey and thus  
76 increase basal resource density (Preisser et al., 2005). Recently, Chang et al. 2019 (under  
77 review) offered another potential explanation. Using a simple consumer-resource model, they

78 showed how the strength of IGP, i.e. number of IG prey consumed by an IG predator per  
79 time, can control whether IGP has a positive or negative effect on basal resource density.  
80 Specifically, their model predicted that basal resource density is a concave up function of the  
81 strength of IGP. When IGP is weak to intermediate in strength, the IG predator increases  
82 more than the decrease of IG prey, such that the summed consumer density increases. In  
83 turn, the basal resource is subjected to the highest predation pressure, and thus has the  
84 lowest density, when IGP is weak to intermediate. Given these results, Chang et al. (under  
85 review) suggested that variation in the strength of IGP among consumer assemblages might  
86 be a plausible explanation for why IGP sometimes has positive, and other times negative  
87 effects on basal resource density in empirical studies. However, this prediction has yet to  
88 be tested with any real biological system. Indeed, no empirical study to our knowledge has  
89 explicitly investigated how the strength of IGP affects basal resource density.

90 In this study, we used a well-developed study system of protozoa consuming bacteria  
91 (Morin, 1999) to run an experiment to test the prediction that basal resource density is  
92 a concave-up function of the strength of IGP—first decreasing as IGP grows from weak to  
93 intermediate strengths, and then increasing as IGP grows from moderate to strong. The  
94 study system was composed of an omnivorous protozoa (the IG predator) that consumed a  
95 strict bacterivore (the IG prey) with which they competed for a common bacterial consortium  
96 (the basal resources). Using this system, we experimentally manipulated the percentage of  
97 the IG prey population that were accessible to the IG predator in order to vary the strength  
98 of IGP.

99 Our paper is organized according to the sequence of our research: First, we ran an  
100 experiment in which we manipulated the strength of IGP in the aforementioned system to  
101 determine how this impacts the density of the basal resource. Subsequently, we fit data from  
102 the experiment to predictions of Chang et al.'s IGP model and realized that while the two  
103 qualitatively agreed, quantitative agreement was poor. Third, suspecting the poor agreement

104 was due to the overly simplistic Type I functional response used in the model, we performed  
105 a second experiment to characterize the functional response of the consumers. Fourth, after  
106 confirming the consumers do, in fact, follow a Type II functional response, we modified the  
107 model accordingly. This resulted in both qualitative and quantitative agreement between  
108 empirical results and model predictions, which allowed us to then use the model as a tool  
109 to deduce the biological mechanism that likely caused resource density to be a concave-up  
110 function of the strength of IGP.

## 111 Method

### 112 Experiment 1 Methods

113 For the experiment, we used a protozoa-bacteria system that has been used previously to  
114 study the stability of food webs (Lawler and Morin, 1993), and to examine how IG prey and  
115 predators coexist (Banerji and Morin, 2014, Morin, 1999). We used this protozoa-bacteria  
116 system to manipulate the strength of IGP, which was accomplished by altering the proportion  
117 of the IG prey population that were available for consumption by the IG predator (hereafter,  
118 availability of IG prey). Manipulating the proportion of IG prey available for consumption is  
119 akin to altering the probability that an IG predator would find an IG prey. The probability of  
120 a predator finding a prey is one of the components of the classic IGP model that determines  
121 the number of prey consumed by a predator per unit of time, i.e. attack rate (Holt and Polis,  
122 1997). We then tested if bacteria (basal resource) density at steady state was a concave-up  
123 function of the strenght of IGP - first decreasing, then increasing as the strenght of IGP  
124 increasd.

125 The focal organisms in the experiment were three bacteria species (*Serratia marcescens*,  
126 *Bacillus cereus*, and *Bacillus subtilis*) that served as the basal resource prey, and two pro-  
127 tozoa, *Colpidium striatum* (IG prey and a strict bacterivore) and *Blepharisma americanum*

128 (IG predator and a omnivore) that served as the consumers (Banerji and Morin, 2014, Morin,  
129 1999). The two protozoa species are known to engage in IGP, but are not known to exhibit  
130 other feeding relationships like cannibalism (Morin, 1999). The focal species were cultured  
131 in  $\sim 300\text{mL}$  experimental bottles that were made from two 240 mL Qorpak glass bottles that  
132 had their bottoms cut off, and which were then glued together with a Bolt Cloth-Nitex mesh  
133 (of varying size, described next) installed in between.

134 The experiment included 6 treatments representing 0%, 20%, 40%, 60%, 80%, and  
135 100% of the IG prey population being made available to the IG predator (Figure 1). To  
136 create the 0% IGP strength treatment, the mesh size of the installed Bolt Cloth-Nitex mesh  
137 was 10-  $\mu\text{m}$ , which was not permeable to both consumers so that no IG prey were available  
138 to the IG predator. The 100% IGP strength treatment was created by replacing the 10- $\mu\text{m}$   
139 mesh with a 250- $\mu\text{m}$  mesh that was permeable to both the IG prey and the IG predator,  
140 such that 100% of the IG prey population was available to the IG predator. For the other  
141 4 treatments, 20- $\mu\text{m}$  mesh that was permeable to the IG prey but not the IG predator  
142 was installed in the experimental bottles. To reassert that the 20- $\mu\text{m}$  mesh was indeed  
143 permeable to the IG prey but not the IG predator, we had used microscope to confirm that  
144 the IG prey could pass through the 20- $\mu\text{m}$  mesh without difficulty but the movement of IG  
145 predator was constrained by the 20- $\mu\text{m}$  mesh. Therefore, the IG prey and the even smaller  
146 bacteria should be homogeneously distributed in the entire experimental unit except in the  
147 0% IGP treatment. The location of the 20- $\mu\text{m}$  mesh was manipulated to divide the entire  
148 experimental bottle into two spaces, a feeding space in which the IG prey was available to  
149 the IG predator, and a refuge space where the IG prey was not available. The ratio of the  
150 feeding space relative to the entire experimental bottle represented the availability of IG prey  
151 to the IG predator, and thus, the strength of IGP. By installing the 20- $\mu\text{m}$  mesh in different  
152 position of the glass bottle, we created 20%, 40%, 60%, 80% IGP strength treatments, each  
153 of which was replicated 5 times.



154 Media for culturing protozoa species in the bottles was created by dissolving 0.07 mg  
155 "protozoan pellets" (Carolina Biological Supply, Burlington, North Carolina) in 1 L sterile  
156 DI water in a 1400-mL flask, after which, the three bacteria species were inoculated. 200 mL  
157 of the media, 2 wheat seeds (one on each side of the experimental unit), and the protozoan  
158 species were then added to the ~300 mL experimental bottle. The experimental bottles  
159 with cultured protozoa were placed on the Thermo Scientific MaxQ 2000 Benchtop Orbital  
160 Shakers to keep the organisms suspended under 60 rounds per minute (rpm), and the shakers  
161 were placed inside a growth chamber where temperature was set to 20-degree Celsius.

162 During the experiment, we monitored the density of the two protozoa species every  
163 other day for 4 weeks. To track protozoa density, 5 mL of media in total was subsampled  
164 from both side of experimental bottle to count the density of both protozoa species once  
165 every other day. After each subsampling, 5 mL fresh sterile media was supplied back to the  
166 experimental bottle to maintain the total volumn. To count protozoan density, 100  $\mu$ L of the  
167 subsampled media was used to count Colpidium striatum (IG prey) density and another 500  
168  $\mu$ L was used to count Blepharisma americanum (IG predator) density under 4X dissecting  
169 microscope.

170 We used the monitoring data for protozoan densities to determine when the experi-  
171 mental units reached steady state with respect to protozoan population density. To assess  
172 steady state, we first calculated the mean density of both protozoa across five consecutive  
173 time points for each IGP treatment. We then gradually moved the five-point time window  
174 forward one time point at a time from hour 34 to hour 468, and divided the mean density of  
175 both protozoa in the present time window by that in the previous time window. The time  
176 window that showed the least change in protozoa density was defined as the steady state.  
177 Experimental systems reached steady state roughly 298 to 468 hours after inoculation of  
178 protozoa (Appendix S1: Figure S1).

179 We next measured bacteria densities at steady state to test the hypothesis that bacterial

180 density first decreases and then increases with IGP strength. To quantify bacterial densities,  
181 we used a Attune<sup>®</sup> Acoustic Focusing Cytometer to count bacteria of each replicate after the  
182 system reached steady state with respect to protozoa density. To prepare samples for the  
183 cytometer, all samples were passed through 20  $\mu\text{m}$  mesh to remove large particles and to avoid  
184 clogging. In addition, if the cell density in a sample was higher than the recommended value  
185 by the Attune<sup>®</sup> Acoustic Focusing Cytometer manual ( $>10^6$  cells per mL), the sample would  
186 be re-run after diluted with Gibco<sup>™</sup> PBS buffers. Finally, we plotted the observed bacteria  
187 density versus IGP strength and fitted a quadratic function to the data to statistically  
188 examine if the extreme values of bacteria density were located within the IGP strength  
189 range (0-100%) of our manipulation. If the coefficient associated with the quadratic term  
190 was significantly less than zero and the constant term was significantly greater than zero, we  
191 concluded that the bacteria density would first decrease and then increase with the increase  
192 of IGP strength. The fitting exercises were done by R 3.5.2 (R Core Team, 2018).

## 193 Experiment 1 Results

194 At steady state, we first fitted a quadratic function to the data to examine if the bacteria  
195 density exhibited a concave-up relationship versus the strength of IGP. The quadratic func-  
196 tion that best fitted to the data had a positive quadratic term (constant = 1; standard error  
197 = 0.14;  $p < 0.01$ ; quadratic term = 1.42, standard error = 0.6,  $p = 0.03$ ;  $R^2 = 0.54$ ). The  
198 internal minimum of the quadratic function occurred at an IGP strength of 37% of the IG  
199 prey population being available to the IG predator (solid line of Figure 2). As the strength  
200 of IGP increased from 0% to ca. 60%, the density of bacteria (basal resources) decreased by  
201 roughly 25% ( $p = 0.05$  for Tukey's Honest Significant Difference test comparing density be-  
202 tween the 0-20% and 40-60% treatments). However, as the strength of IGP further increased  
203 to 80% and 100%, bacterial density increased by 36% relative to the 0% IGP treatment ( $p$   
204 = 0.02 for Tukey's Honest Significant Difference test comparing density between 0-20% and  
205 80-100% treatments; Figure 2). Qualitatively, these experimental data match the theoretical

206 prediction from Chang et al. 2019 (under review) that the strength of IGP first decreases,  
207 and then increases basal resource density.

208 Although the empirical data on bacteria density at steady state from experiment 1  
209 qualitatively matched our a priori prediction, the empirical data (Figure 2 solid dots) no-  
210 tably diverged from the model output of Chang et al. 2019 (under review), from which our  
211 predictions were derived (Figure 2 long dashed line). The difference between model predic-  
212 tions and empirical data (sum of square = 0.697) was greater than the total sums of squares  
213 of the data (0.365), which means the model prediction from Chang et al. 2019 (under review)  
214 was a poorer fit to the empirical data than the grand mean of the data. To improve the  
215 match between model predictions and empirical results, our first attempt was to parame-  
216 terized Chang et al's with values from literature (Table 1). Unfortunately, parameterizing  
217 Chang et al's model with literature values further increase the difference between model  
218 predictions and empirical data (sum of square = 65.05; dotted line in Figure 2). Because of  
219 the poor fit, we decided to pursue additional experimental work and model revisions that  
220 would achieve a better match of empirical data and theoretical predictions.

## 221 Experiment 2 Methods

222 We suspected that the most likely reason why bacterial densities measured in Experiment 1  
223 did not quantitatively match model predictions of Chang et al. 2019 (under review) was that  
224 the authors used an overly simplistic Type I functional response to model all consumption  
225 terms. In contrast to the simple Type I, some authors have suggested the consumption  
226 of bacteria by both Colpidium and Blepharisma may be better approximated by Type II  
227 functional response (Laybourn and Stewart, 1975). To determine the type of functional  
228 response exhibited by the two protozoa species, we performed an additional experiment to  
229 quantify the functional response curve describing the consumption of IG prey by IG predator.  
230 This additional experiment was run in 60 mm (diameter) x 15 mm (height) Fisherbrand™

231 Petri Dishes with Clear Lid. We set up 3 replicate units for each of 10 treatments representing  
232 5 levels of IG prey density with one IG predator individual, and the same 5 levels of IG prey  
233 density without IG predator. The five levels of IG prey density were created by mixing 2, 4,  
234 6, 8, and 10 mL of media with Colpidium (IG prey) with 8, 6, 4, 2, 0 mL of protozoa-free  
235 medium. The average density of IG prey of the 5 levels in the beginning was 4.11, 9.89, 15.11,  
236 20, 29.78 individual/mL. The Petri dishes were then placed also on the Thermo Scientific  
237 MaxQ 2000 Benchtop Orbital Shakers rotating at 60 rpm in the same 20-degree Celsius  
238 growth chamber. After 24 hours, we recorded the density changes of IG prey in treatments  
239 with and without IG predator. The differences between the density changes in treatments  
240 with and without IG predator were the number of IG prey consumed per IG predator per  
241 day.

## 242 Experiment 2 - Results

243 By plotting the number of IG prey consumed per IG predator per day against the initial  
244 IG prey density, we found that a Type II saturating functional response (dashed line in  
245 Figure 3a.;  $R^2 = 0.91$ ) was a better explanation of the intra-guild predation than Type I  
246 linear functional response ( $p < 0.01$ ). From the Type II saturating function, the IGP attack  
247 rate and handling time were estimated to be 0.39 and 0.36, respectively. Given these results,  
248 we decided the next step was to modify all consumption terms in Chang et al. 2019 (under  
249 review) with a Type II functional response, and then parameterized the revised model with  
250 results of Experiment 2 (the IGP attack rate and handling time) or with values published  
251 in the literature.

## 252 A revised IGP model with Type II functional response

253 Using the same general model structure as Chang et al. 2019 (under review), we used the  
254 following four equations to describe the population dynamics of two basal resources as well  
255 as IG prey and predator.

$$\frac{dR_1}{dt} = r_1 R_1 \left(1 - \frac{R_1}{K_1}\right) - \left[ \frac{c_1 s R_1}{1 + h_1 c_1 s R_1 + h_1 c_1 (1-s) R_2} \right] Z_1 - \left[ \frac{c_2 (1-s) R_1}{1 + h_2 c_2 (1-s) R_1 + h_2 c_2 s R_2 + h_3 c_3 \alpha Z_1} \right] Z_2, \quad (1)$$

$$\frac{dR_2}{dt} = r_2 R_2 \left(1 - \frac{R_2}{K_2}\right) - \left[ \frac{c_1 (1-s) R_2}{1 + h_1 c_1 s R_1 + h_1 c_1 (1-s) R_2} \right] Z_1 - \left[ \frac{c_2 s R_2}{1 + h_2 c_2 (1-s) R_1 + h_2 c_2 s R_2 + h_3 c_3 \alpha Z_1} \right] Z_2, \quad (2)$$

$$\frac{dZ_1}{dt} = \left[ \frac{e_1 c_1 s R_1 + e_1 c_1 (1-s) R_2}{1 + h_1 c_1 s R_1 + h_1 c_1 (1-s) R_2} \right] Z_1 - \left[ \frac{c_3 \alpha Z_1}{1 + h_2 c_2 (1-s) R_1 + h_2 c_2 s R_2 + h_3 c_3 \alpha Z_1} \right] Z_2 - m Z_1, \quad (3)$$

$$\frac{dZ_2}{dt} = \left[ \frac{e_2 c_2 (1-s) R_1 + e_2 c_2 s R_2 + e_3 c_3 \alpha Z_1}{1 + h_2 c_2 (1-s) R_1 + h_2 c_2 s R_2 + h_3 c_3 \alpha Z_1} \right] Z_2 - m Z_2, \quad (4)$$

256 In accordance with Chang et. al. 2019 (under review), the dynamics of bacteria species  
 257 ( $R_1$  and  $R_2$ ), IG prey ( $Z_1$ ) and predator ( $Z_2$ ) were described by equation 1 to 4 respectively.  
 258 The two basal resources grew logistically with intrinsic growth rates  $r_1$  and  $r_2$ , as well as  
 259 carrying capacities  $K_1$  and  $K_2$ . Both basal resources were consumed by IG prey ( $Z_1$ ) and  
 260 IG predator ( $Z_2$ ) following a Type I functional response with attack rate ( $c_i$ ), where  $i = 1$   
 261 or 2 indicating IG prey or IG predator, respectively. Following Chang et al. 2019 (under  
 262 review), the parameter  $s$ , ranging from 0.5 to 1, was designed to manipulate the degree of  
 263 resource partitioning among IG prey and predator. When  $s = 0.5$ , both consumers become  
 264 complete generalists consuming equally on both resources. When  $s = 1$  the IG prey ( $Z_1$ )  
 265 was a complete specialists consuming  $R_1$ , and the predator ( $Z_2$ ) was completely specialized  
 266 on  $R_1$ . The dynamics of IG prey ( $Z_1$ ) and IG predator ( $Z_2$ ) were described by equation 3  
 267 and 4. Growth rate of IG prey and IG predator was determined by the consumption terms,  
 268 which now followed a Type II functional response, multiplied the assimilation efficiency ( $e_i$ ).  
 269 In addition, the IG predator also consumed the IG prey, i.e. the intra-guild predation,  
 270 following also a Type II functional response with the IGP attack rate ( $c_3$ ), handling time

271 ( $h_3$ ), assimilation efficiency ( $e_3$ ), and the parameter  $\alpha$  describing the availability of IG prey  
272 to IG predator. The  $\alpha$  is the strength of IGP that is the focus of this study. Finally, both  
273 IG prey and predator had a density independent mortality ( $m_1$  and  $m_2$ ).

274 To generate predictions from the Type II model (equations 1-4), we parameterized the  
275 model with values from experiment 2 (the IGP attack rate and handling time; Figure 3a)  
276 and the published literature (Type II model column of Table 1), with exception of two  
277 parameters: the assimilation efficiency from IG prey to IG predator ( $e_3$ ) and the degree  
278 of resource partitioning ( $s$ ). We estimated the assimilation efficiency from IG prey to IG  
279 predator ( $e_3$ ) and the degree of resource partitioning ( $s$ ) by searching for the combination  
280 of two parameters that yielded the least difference between model predictions and empirical  
281 data. To estimate the difference between model predictions and empirical data, we first  
282 plotted the summed density of bacteria as well as the density of IG prey and predator at  
283 steady state against IGP strength ( $\alpha$ ), i.e. 0%, 20%, 40%, 60%, 80%, 100% of IG prey  
284 available to IG predator. These empirical patterns were overlaid with predictions from the  
285 Type II model to calculate the residual sum of square of the model predictions. Note that  
286 the model predictions were calculated from the long-term average (last 2000 time steps of  
287 the 10000 simulated time steps) of two resources,  $Z_1$  and  $Z_2$  because the model appears to  
288 exhibit limit cycle behaviors (Appendix S1, and Appendix S1: Figure S2). By doing this,  
289 we found the best parameter value combination of assimilation efficiency ( $e_3$ ) and degree of  
290 resource partitioning ( $s$ ) to be 40% and 96% respectively (Figure 3b). The model predictions  
291 was generated with the aid of Mathematica 11.1 (Wolfram Research, Inc., 2017) and the  
292 difference between model predictions and empirical data was done by R 3.5.2 (R Core Team,  
293 2018).

294 The revised IGP model with a Type II functional response explained 67%, 68%, and  
295 66% of the variance for summed bacterial density, as well as the density of the IG prey  
296 and the density of the IG predator at steady state in experiment 1, respectively (Figure 4).

297 The model with a Type II functional response more appropriately captured the threshold of  
298 bacteria density (40-60% availability of IG prey population to IG predator) beyond which  
299 bacteria density started to increase with the strength of IGP (Figure 4a). For the IG prey  
300 density at steady state, the Type II model also predicted the monotonic decrease from 0%  
301 to 100% IGP strength (the solid line in Figure 4b) that was observed in the experiment  
302 (solid dots in Figure 4b). Finally, for the IG predator density, the Type II model and the  
303 experimental data suggested that IGP strength actually increased and then leveled off  
304 (Figure 4c). Given the improved match between empirical data and predictions from the  
305 Type II model, we can infer that the decrease and then increase of bacterial density at steady  
306 state (Figure 2 and Figure 4a) resulted from the saturating Type II functional response of  
307 consumers involved in IGP (explained further in the discussion).

## 308 Discussion

309 Our experimental results showed that, at the steady state, the density of basal prey resource  
310 (bacteria) was a concave-up function of intra-guild predation - first decreasing, and then  
311 increasing as the strength of IGP increased (Figure 2 and Figure 4a). This finding supports  
312 the theoretical prediction from Chang et al. 2019 (under review) that intra-guild predation  
313 strength would first decrease and then increases the density of basal resource. This finding  
314 could help explain why some empirical studies have demonstrated negative effects of IGP  
315 on basal resource density, whereas others have shown the opposite (Vance-Chalcraft et al.,  
316 2007).

317 The observed population dynamics of bacteria, IG prey and predator were explained  
318 reasonably well ( $\sim 70\%$ ) by an IGP model that was modified to have a Type II functional re-  
319 sponse (solid lines in Figure 4). That model was partly parameterized with literature values,  
320 and partly parameterized by our own experimental work. The Type II model helped identify  
321 potential biological mechanisms that might underlie the patterns observed in experiment 1.

322 When the IGP strength increased from 0% to more moderate levels (e.g. 40-60%), the IG  
323 predator density increased due to moderate consumption of the IG prey. The summed con-  
324 sumer density was, therefore, higher such that bacteria density was suppressed to the lowest  
325 level. When the strength of IGP increased further, the IG predator started to be satiated by  
326 the IG prey such that the IG predator stopped consuming bacteria and the IG prey density  
327 reached the lowest level. Consequently, the lowest level of IG prey density and satiated IG  
328 predator resulted in the lowest bacterial consumption and thus highest bacterial resource  
329 density at equilibrium (Figure 2 and Figure 4).

330 The fit of the Type II model to data from the experiment is a bit of a double-edged  
331 sword. On one hand, the model did appear to qualitatively capture the concave-up rela-  
332 tionship between the strength of IGP and the density of the bacterial resources, which was  
333 our original prediction based on previous theoretical work. In addition, the model was a  
334 reasonable fit to this concave up relationship, explaining approximately 70% of the variation  
335 in empirical data. On the other hand, the Type II model that we developed failed to mimic  
336 certain aspects of the temporal dynamics of the experiment. Indeed, population densities  
337 of the IG prey and predator reached a steady state over the time-frame of the experiment  
338 (Appendix S1: Figure S1), whereas the Type II model was characterized by limit cycles  
339 (Appendix S1: Figure S2). While the mean values of the model matched those from the  
340 experiment, there clearly was a mismatch in temporal dynamics that suggests certain aspects  
341 of biology are still missing from model. We speculate that the most likely mismatch lies in  
342 the accuracy of the measures of bacterial growth rate and/or conversion efficiencies from  
343 bacteria to protozoa. A reduction in these parameter values could potentially dampen the  
344 limit cycles. Despite clear limitations of the model, we feel that an imperfect quantitative  
345 description of the experimental system is better than none at all.

346 Though the assimilation efficiency from IG prey to IG predator has not been empirically  
347 measured, it is possible to infer a value using size spectrum theory (Kerr, 1974, Sheldon



348 et al., 1977). According to the size spectrum theory, an individual's consumption rate and  
349 volumetric search rate should scale with body size as a power law with scaling exponents  
350 of 0.75 and 0.80, respectively (Andersen and Beyer, 2006). Given those scaling exponents  
351 and the predator-prey body size ratio in our study (Long and Morin, 2005), the assimilation  
352 efficiency is predicted to be 41.7% (Andersen et al., 2009), which is similar to our fitted  
353 estimate (40%).

354 While the value for assimilation efficiency is consistent with other lines of evidence, we  
355 are somewhat skeptical of the value we obtained for resource partitioning between the IG  
356 prey and IG predator. Our model suggested there was a 96% separation in food items con-  
357 sumed by the IG predator and the IG prey. While the two species certainly exhibit resource  
358 partitioning, biologically, these two species are both filter feeders with relatively similar cil-  
359 iary structures (Fenchel, 1980). Because complete partitioning of bacterial resources seems  
360 unlikely, this particular parameter needs to be verified or refined with additional experi-  
361 ments.

362 In addition to the two fitted parameters ( $e_3$  and  $s$ ), attack rate ( $c_2$ ) and assimilation  
363 efficiency ( $e_2$ ) of *Blepharisma* on bacteria were set to be the same as that of *Colpidium*.  
364 We made this decision because both species were both filter feeders (Verni and Gualtieri,  
365 1997, Thurman et al., 2010) and they fed on the same food resources in our experiments  
366 although these two parameters may differ among the two species. However, altering the  
367 value of these two parameters did not appear to qualitatively change the match between  
368 model predictions and empirical results (Appendix S2). After altering the value of attack  
369 rate ( $c_2$ ) and assimilation efficiency ( $e_2$ ) we still found the density of basal resources first  
370 decrease and then increase with intra-guild predation. Consequently, the value of attack  
371 rate ( $c_2$ ) and assimilation efficiency ( $e_2$ ) of *Blepharisma* on bacteria can differ from that of  
372 *Colpidium* but such difference should not change our conclusions.

373 Our experiment suffers all the caveats that are typical of microcosms experiments

374 (Briggs and Borer, 2005). These caveats include small temporal and spatial scale, restricted  
375 environmental variability and species interactions, as well as lack of natural trophic struc-  
376 tures. In addition, our experimental setup differs from the model, although we have done  
377 our best efforts to design and parameterize the model to match the empirical experiments.  
378 In the experiments, there were three bacteria species, but there were two resources in the  
379 theoretical model. We can only speculate how having the third resource in the theoretical  
380 model might influence the model agreement to our empirical work. One of our speculations  
381 was that including the third resource only increased the degree of resource partitioning if the  
382 third resource was mainly consumed by IG prey or predator. Since the degree of resource  
383 partitioning was already estimated to be very high (96%), having the third resource might  
384 not influence the match between theoretical model and empirical results. On the other hand,  
385 addition of the third resource might decrease the degree of resource partitioning if the third  
386 resource was evenly consumed by IG prey and predator or the third resource might change  
387 the population density of IG prey or predator. In such scenario, the match between theo-  
388 retical predictions and empirical results would be deteriorated. We can't discern the actual  
389 impacts of having the third resource before more research has been done. Nevertheless, we  
390 would point out that our goal in this study is not to extend results to real aquatic ecosystems.  
391 Rather, the point of a microcosm experiment like this one is to test a specific theoretical  
392 prediction in highly controlled environment. Now that we have done so, the next obvious  
393 step is to similarly test the prediction using observational studies or experiments in more  
394 natural systems.

395 To our knowledge, this is the first empirical study to explicitly investigate how the  
396 strength of intra-guide predation impacts basal resource density. Our finding that IGP can  
397 decrease or increase basal resource density depending on its strength runs counter to the  
398 conventional thinking that IGP always interferes the ability of consumers to control basal  
399 resources and, in turn, always increases basal resource density. Our study suggests that  
400 the impacts of IGP could be more complex than previously expected. Therefore, if we are

401 to better understand what controls the consumption of basal resources, we may need to  
402 explicitly quantify the strength of intra-guild predation.

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## References

- Andersen, K. H. and J. E. Beyer. 2006. Asymptotic Size Determines Species Abundance in the Marine Size Spectrum. *The American Naturalist*, 168:54–61. URL <https://doi.org/10.1086/504849>.
- Andersen, K. H., J. E. Beyer, and P. Lundberg. 2009. Trophic and individual efficiencies of size-structured communities. *Proceedings of the Royal Society B: Biological Sciences*, 276:109–114. URL <http://rspb.royalsocietypublishing.org/cgi/doi/10.1098/rspb.2008.0951>.
- Arim, M. and P. A. Marquet. 2004. Intraguild predation: A widespread interaction related to species biology. *Ecology Letters*, 7:557–564.
- Banerji, A. and P. J. Morin. 2014. Trait-mediated apparent competition in an intraguild predator–prey system. *Oikos*, 123:567–574. URL <https://doi.org/10.1111/j.1600-0706.2013.00937.x>.
- Barnes, A. D., M. Jochum, J. S. Lefcheck, N. Eisenhauer, C. Scherber, M. I. O’Connor, P. de Ruiter, and U. Brose. 2018. Energy Flux: The Link between Multitrophic Biodiversity and Ecosystem Functioning. *Trends in Ecology and Evolution*, 33:186–197. URL <https://www.sciencedirect.com/science/article/pii/S0169534717303257?via=ihub>.
- Briggs, C. J. and E. T. Borer. 2005. Why short-term experiments may not allow long-term predictions about intraguild predation. *Ecological Applications*, 15:1111–1117.
- Daugherty, M. P., J. P. Harmon, and C. J. Briggs. 2007. Trophic supplements to intraguild predation. *Oikos*, 116:662–677.
- Duffy, J. E. and A. M. Harvilicz. 2001. Species-specific impacts of grazing amphipods in

434 an eelgrass-bed community. *Marine Ecology Progress Series*, 223:201–211. URL <https://www.int-res.com/abstracts/meps/v223/p201-211/>.

436 Dunne, J., R. J. Williams, and N. D. Martinez. 2004. Network structure and robustness of  
437 marine food webs. *Marine Ecology Progress Series*, 273:291–302.

438 Fedrigo, G. V., E. M. Campoy, G. Di Venanzio, M. I. Colombo, and E. García Vescovi. 2011.  
439 *Serratia marcescens* is able to survive and proliferate in autophagic-like vacuoles inside  
440 non-phagocytic cells. *PLoS ONE*, 6:e24054.

441 Fenchel, T. 1980. Suspension feeding in ciliated protozoa: Feeding rates and their ecological  
442 significance. *Microbial Ecology*, 6:13–25. URL [http://link.springer.com/10.1007/  
443 BF02020371](http://link.springer.com/10.1007/BF02020371).

444 Finke, D. L. and W. E. Snyder. 2008. Niche Increases Resource Partitioning by Diverse  
445 Communities Exploitation. *Science*, 321:1488–1490.

446 Hart, D. R. 2002. Intraguild predation, invertebrate predators, and trophic cascades in lake  
447 food webs. *Journal of Theoretical Biology*, 218:111–128.

448 Holt, R. D. and G. A. Polis. 1997. A Theoretical Framework for Intraguild Predation. *The*  
449 *American Naturalist*, 149:745–764. URL <https://doi.org/10.1086/286018>.

450 Hutchinson, G. E. 1957. Concluding remarks. *Cold Spring Harbor Symposia on Quantitative*  
451 *Biology*, 22:415–427.

452 Hutchinson, G. E. 1961. The Paradox of the Plankton. *The American Naturalist*, 95:137–145.  
453 URL <http://www.jstor.org/stab>.

454 Kerr, S. R. 1974. Theory of Size Distribution in Ecological Communities. *Journal of the*  
455 *Fisheries Research Board of Canada*, 31:1859–1862. URL [https://doi.org/10.1139/  
456 f74-241](https://doi.org/10.1139/f74-241).

- 457 Kuijper, L. D. J., B. W. Kooi, C. Zonneveld, and S. A. L. M. Kooijman. 2003. Omnivory  
458 and food web dynamics. *Ecological Modelling*, 163:19–32.
- 459 Lawler, S. P. and P. J. Morin. 1993. Food web architecture and population dynamics in  
460 laboratory microcosms of protists. *The American naturalist*, 141:675–686.
- 461 Laybourn, J. E. M. and J. M. Stewart. 1975. Studies on Consumption and Growth in the  
462 Ciliate *Colpidium campylum* Stokes. *Journal of Animal Ecology*, 44:165–174.
- 463 Long, Z. T. and P. J. Morin. 2005. Effects of organism size and community composition on  
464 ecosystem functioning. *Ecology Letters*, 8:1271–1282.
- 465 Losey, J. E. and R. F. Denno. 1998. Positive predator–predator interactions: enhanced  
466 predation rates and synergistic suppression of aphid populations. *Ecology*, 79:2143–2152.  
467 URL [https://doi.org/10.1890/0012-9658\(1998\)079\[2143:PPPIEP\]2.0.CO;2](https://doi.org/10.1890/0012-9658(1998)079[2143:PPPIEP]2.0.CO;2).
- 468 Losey, J. E. and R. F. Denno. 1999. Factors facilitating synergistic predation: the central  
469 role of synchrony. *Ecological Applications*, 9:378–386. URL [https://doi.org/10.1890/1051-0761\(1999\)009\[0378:FFSPTC\]2.0.CO;2](https://doi.org/10.1890/1051-0761(1999)009[0378:FFSPTC]2.0.CO;2)  
470 <http://0.0.0.2>.
- 471 MacArthur, R. H. 1958. Population ecology of some warblers of northeastern coniferous  
472 forests. *Ecology*, 39:599–619. URL <http://www.jstor.org/stable/1931600>.
- 473 Morin, P. 1999. Productivity, intraguild predation, and population dynamics in experimental  
474 food webs. *Ecology*, 80:752–760. URL <http://www.jstor.org/stable/177014>.
- 475 Polis, G. A. and R. D. Holt. 1992. Intraguild predation: The dynamics of complex  
476 trophic interactions. *Trends in Ecology and Evolution*, 7:151–154. URL <http://www.sciencedirect.com/science/article/pii/016953479290208S>.
- 477
- 478 Polis, G. A., C. A. Myers, and R. D. Holt. 1989. The ecology and evolution of intraguild  
479 predation: potential competitors that eat each other. *Annual Review of Ecology and*  
480 *Systematics*, 20:297–330.

481 Preisser, E. L., D. I. Bolnick, and M. F. Benard. 2005. Scared to death? the effects of  
482 intimidation and consumption in predator-prey interactions. *Ecology*, 86:501–509. URL  
483 <https://esajournals.onlinelibrary.wiley.com/doi/abs/10.1890/04-0719>.

484 R Core Team. 2018. R: A Language and Environment for Statistical Computing. URL  
485 <https://www.r-project.org/>.

486 Ratkowsky, D. A., J. Olley, T. A. McMeekin, and A. Ball. 1982. Relationship between  
487 temperature and growth rate of bacterial cultures. *Journal of Bacteriology*, 149:1–5.

488 Rosenheim, J. A. and J. P. Harmon. 2006. The Influence of Intraguild Predation on the  
489 Suppression of a Shared Prey Population: An Empirical Reassessment. In J. Brodeur  
490 and G. Boivin, editors, *Trophic and Guild in Biological Interactions Control*, pages 1–20.  
491 Springer Netherlands, Dordrecht. URL [https://doi.org/10.1007/1-4020-4767-3\\_{\\_}1](https://doi.org/10.1007/1-4020-4767-3_{_}1).

492 Rudolf, H. W. V. 2007. The interaction of cannibalism and omnivory: Consequences for  
493 community dynamics. *Ecology*, 88:2697–2705.

494 Schluter, D. 1993. Adaptive radiation in sticklebacks: size, shape, and habitat use efficiency.  
495 *Ecology*, 74:699–709. URL <http://www.jstor.org/stable/1940797>.

496 Sheldon, R. W., W. H. Sutcliffe Jr., and M. A. Paranjape. 1977. Structure of Pelagic Food  
497 Chain and Relationship Between Plankton and Fish Production. *Journal of the Fisheries*  
498 *Research Board of Canada*, 34:2344–2353. URL <https://doi.org/10.1139/f77-314>.

499 Sih, A., G. Englund, and D. Wooster. 1998. Emergent impacts of multiple predators on prey.  
500 *Trends in Ecology and Evolution*, 13:350–355.

501 Thompson, R. M., M. Hemberg, B. M. Starzomski, and J. B. Shurin. 2007. Trophic levels  
502 and trophic tangles: the prevalence of omnivory in real food webs. *Ecology*, 88:612–617.  
503 URL <https://doi.org/10.1890/05-1454>.

- 504 Thurman, J., J. D. Parry, P. J. Hill, and J. Laybourn-Parry. 2010. The filter-feeding ciliates  
505 colpidium striatum and tetrahymena pyriformis display selective feeding behaviours in  
506 the presence of mixed, equally-sized, bacterial prey. *Protist*, 161:577 – 588. URL [http:  
507 //www.sciencedirect.com/science/article/pii/S1434461010000234](http://www.sciencedirect.com/science/article/pii/S1434461010000234).
- 508 Vance-Chalcraft, H. D., J. A. Rosenheim, J. R. Vonesh, C. W. Osenberg, and A. Sih. 2007.  
509 The influence of intraguild predation on prey suppression and prey release: A meta-  
510 analysis. *Ecology*, 88:2689–2696.
- 511 Verni, F. and P. Gualtieri. 1997. Feeding behaviour in ciliated protists. *Micron*, 28:487 – 504.  
512 URL <http://www.sciencedirect.com/science/article/pii/S0968432897000280>.
- 513 Wolfram Research, Inc. 2017. *Mathematica*, Version 11.2. Champaign, Illinois, USA.



Table 1 Parameter values used for model with Type I functional response in Chang et al. 2019 (under review) and model with Type II functional response in this study.

Parameter	Value			Source for Type II model
	Chang et al.'s Type I model	Chang et al.'s Type I model (with literature values)	Type II model	
– Bacteria per capita growth rate ( $r_i$ ; 1/day)	2.5	1.72	1.72	Ratkowsky et al. (1982), Fedrigo et al. (2011)
– Bacteria carrying capacity ( $K_i$ ; ind./mL)	$50 \times 10^5$	$8.14 \times 10^{5\dagger}$	$8.14 \times 10^{5\dagger}$	Empirically measured
– Attack rate of Colpidium (IG prey) on bacteria ( $c_1$ ; 1/day · consumer)	1	1.25	1.25	Laybourn and Stewart (1975)
– Handling time of Colpidium (IG prey) on bacteria ( $h_1$ ; day · consumer/resource)	N.A.	N.A.	$0.08 \times 10^5$	Laybourn and Stewart (1975)
– Assimilation efficiency from bacteria to Colpidium (IG prey) ( $e_1$ ; %)	0.3	0.11	0.11	Laybourn and Stewart (1975)
– Attack rate of Blepharisma (IG predator) on bacteria ( $c_2$ ; 1/day · consumer)	0.7	1.25	1.25	Same as Colpidium but see Appendix S2
– Handling time of Blepharisma (IG predator) on bacteria ( $h_2$ ; day · consumer/resource)	N.A.	N.A.	$0.8 \times 10^{5\dagger}$	Laybourn and Stewart (1975), Fenchel (1980)
– Assimilation efficiency from bacteria to Blepharisma (IG predator) ( $e_2$ ; %)	0.3	0.11	0.11	Same as Colpidium but see Appendix S2
– IGP attack rate ( $c_3$ ; 1/day · consumer)	1	0.39	0.39	Empirically measured
– IGP Handling time ( $h_3$ ; day / consumer · resource)	N.A.	N.A.	0.36	Empirically measured
– IGP assimilation efficiency ( $e_3$ ; day / consumer · resource)	1	0.4	0.4	Fig. 3b
– Degree of resource partitioning ( $s$ )	0.75	0.96	0.96	Fig. 3b
– Density independent mortality ( $m$ )	1	0.1	0.1	Empirically measured

515 † This bacterial density should be high enough for the IG predator to exhibit non-significantly

516 different bacterial consumption rate among IGP treatments.

517 ‡This value was estimated from the fact that the maximum food uptake rate (which should  
518 be the inverse of handling time) of Colpidium was 10 times higher than that of Blepharisma  
519 (Fenchel, 1980).

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520 Figure captions

521 Figure 1

522 Conceptual figure showing the simplified food web structure on the left and the experimen-  
523 tal design of this study on the left. On the right, the rounded squares represent the six  
524 experimental treatments (0, 20, 40, 60, 80 and 100% IGP strength). The B and C in the  
525 rounded square are the initial of the IG predator (Blepharisma) and IG prey (Colpedium)  
526 species used in this study. The solid line in the first square from the left indicates that IG  
527 prey is not accessible to the IG predator (0% IGP). The dashed lines of the central four  
528 squares represent the 20  $\mu\text{m}$  mesh that is permeable to IG prey but not IG predator. The  
529 location of the 20  $\mu\text{m}$  mesh manipulates the percentage of IG prey that is accessible to IG  
530 predator, 20, 40, 60, and 80% IGP strength. The long-dashed line in the rightmost square  
531 represents the 250  $\mu\text{m}$  mesh that is permeable to both IG prey and predator so that 100%  
532 IGP is allowed.

533 Figure 2

534 Mean population density of bacteria when the experimental system reached steady state  
535 with respect to protozoa density (roughly hour 298 to 468) in different IGP strength  
536 treatments. The error bars represent the standard error of the mean. The 80% and 100%  
537 treatments have significantly higher bacteria density ( $p < 0.01$ ). The three different lines  
538 represent (1) the quadratic function that best fits to the data (solid line with its standard  
539 error), (2) the predictions from the model of Chang et al. 2019 (long-dashed line) and  
540 (3) the model prediction when re-parameterizing Chang et al.'s model (Type I model) with  
541 parameter values from literature. Note that both the long dashed line and the dotted line  
542 are both from Chang et al.'s model but parameterized with different sets of values, which  
543 poorly fit the data (see text).

544 Figure 3

545 Panel a. shows the Type II functional response of intra-guild predation, i.e. number of IG

546 prey consumed against IG prey versus IG predator ratio. The two parameters describing  
547 the Type II functional response, IGP attack rate and handling time of *Blepharisma* (IG  
548 predator) on *Colpidium* (IG prey), are estimated to be 0.39 and 0.36 respectively. Panel  
549 b. shows the total residual sum of square of the model with different combinations of the  
550 two unknown parameters, the degree of resource partitioning between IG prey and predator  
551 as well as the assimilation efficiency from IG prey to IG predator. At a given combination  
552 of the two parameter values, the normalized total residual sum of square of the model for  
553 bacteria, IG prey and predator density is calculated and represented by the color of the tile.  
554 White space represents the combination that the model results in higher normalized total  
555 residual sum of squared than just the average across intra-guild predation treatments (a null  
556 model). The model has the lowest total residual sum of square when the degree of resource  
557 partitioning is 0.96 and the assimilation efficiency is 40%.

558 Figure 4

559 Mean population density of bacteria (panel a), IG prey (panel b) and IG predator (panel  
560 c), when the experimental system reached steady state with respect to protozoa density  
561 (roughly hour 298 to 468) in different IGP strength treatments. The error bars represent  
562 the standard error of the mean. The solid lines represent the predictions from the model  
563 with Type II functional response (Type II model). The solid lines explain 66%, 68.08%, and  
564 67.45% of the variance across the 6 IGP strength treatments (0%-100%) for IG predator, IG  
565 prey and bacteria density respectively.

Figure 1

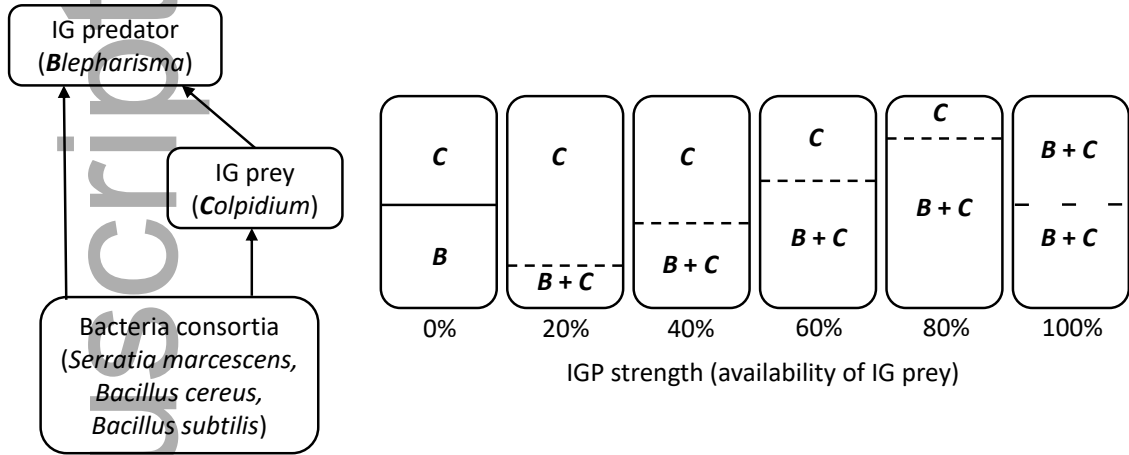


Figure 2

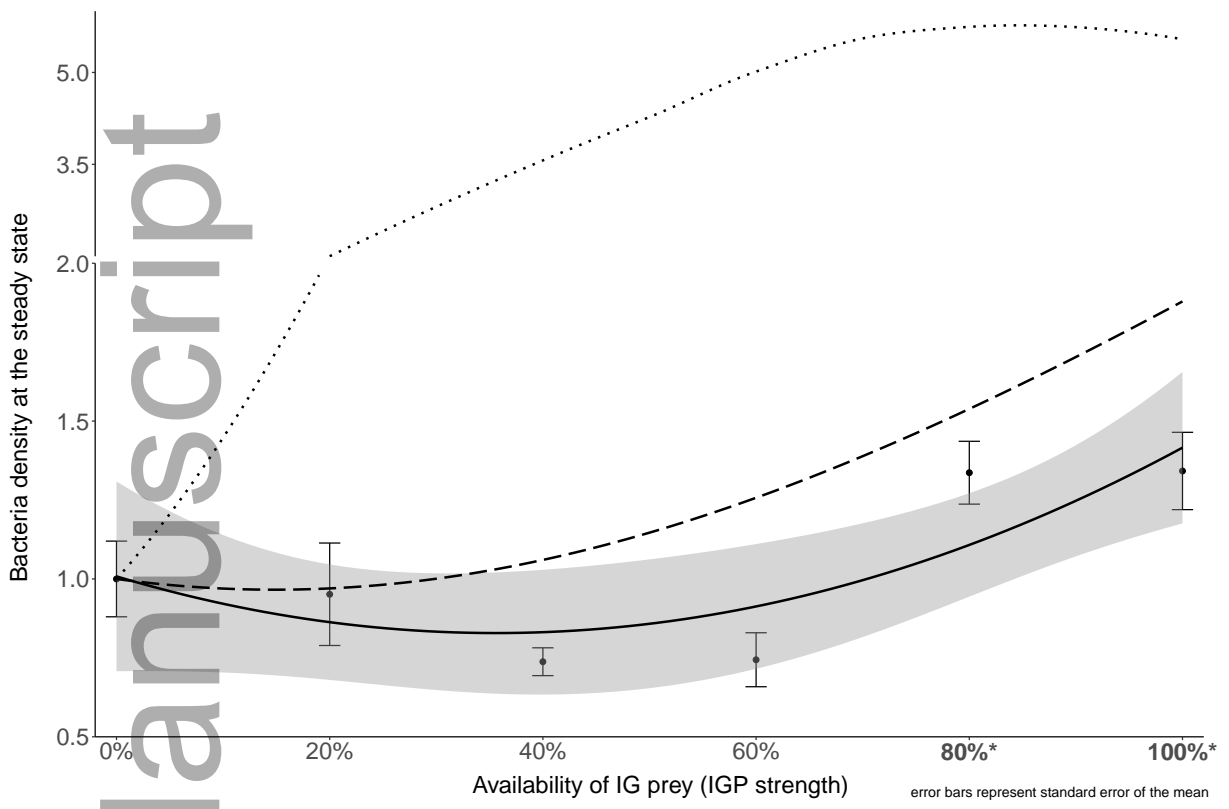


Figure 3

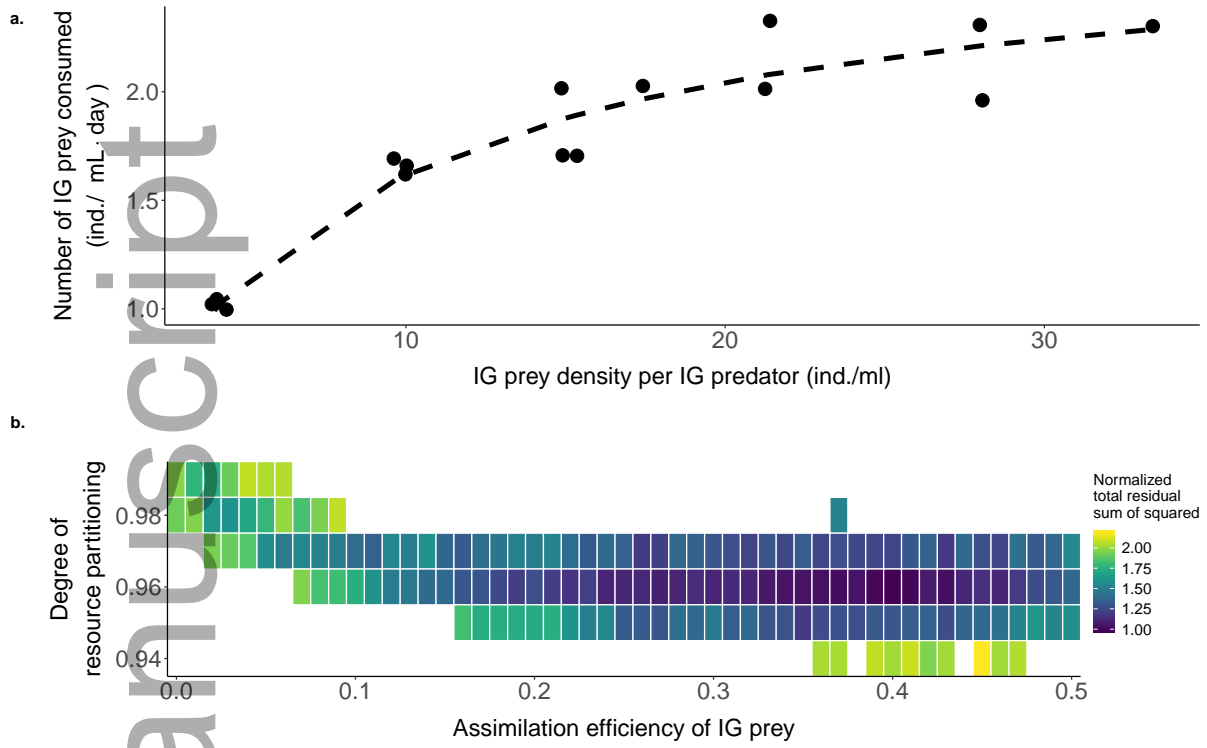
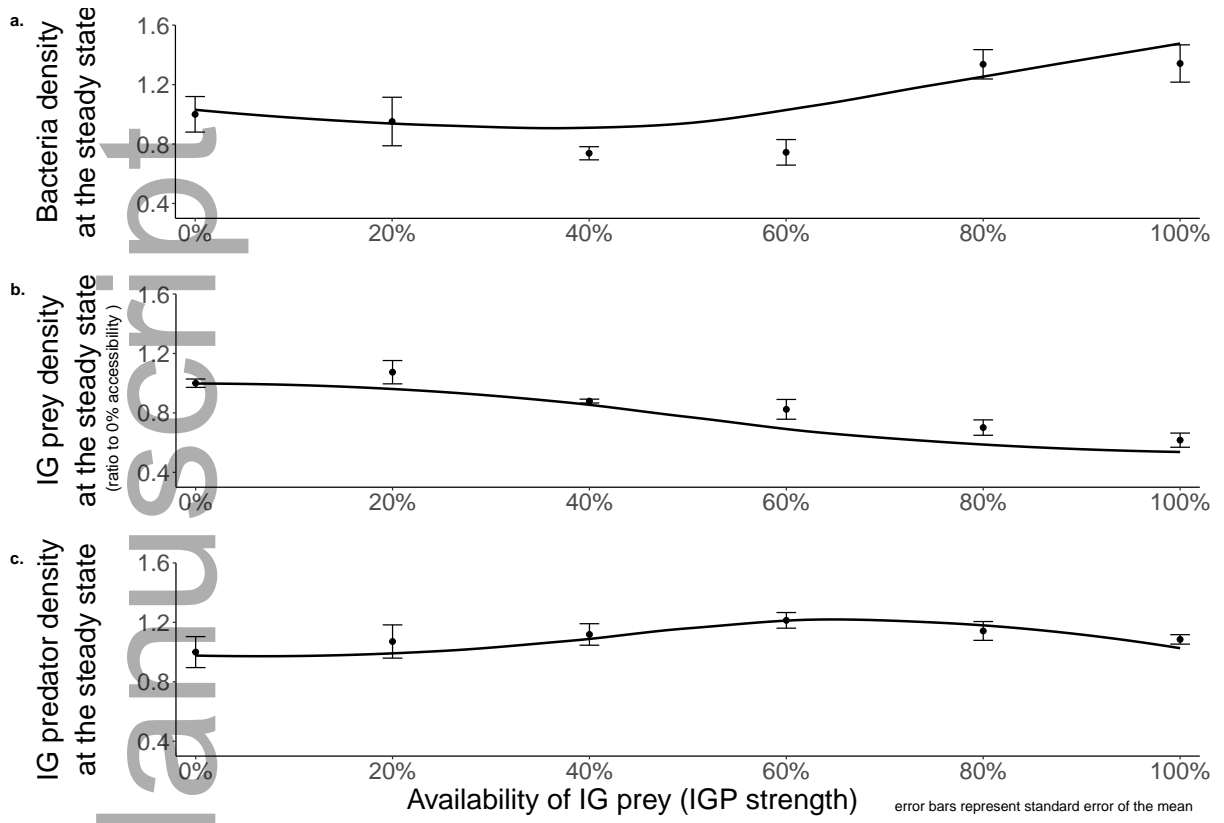
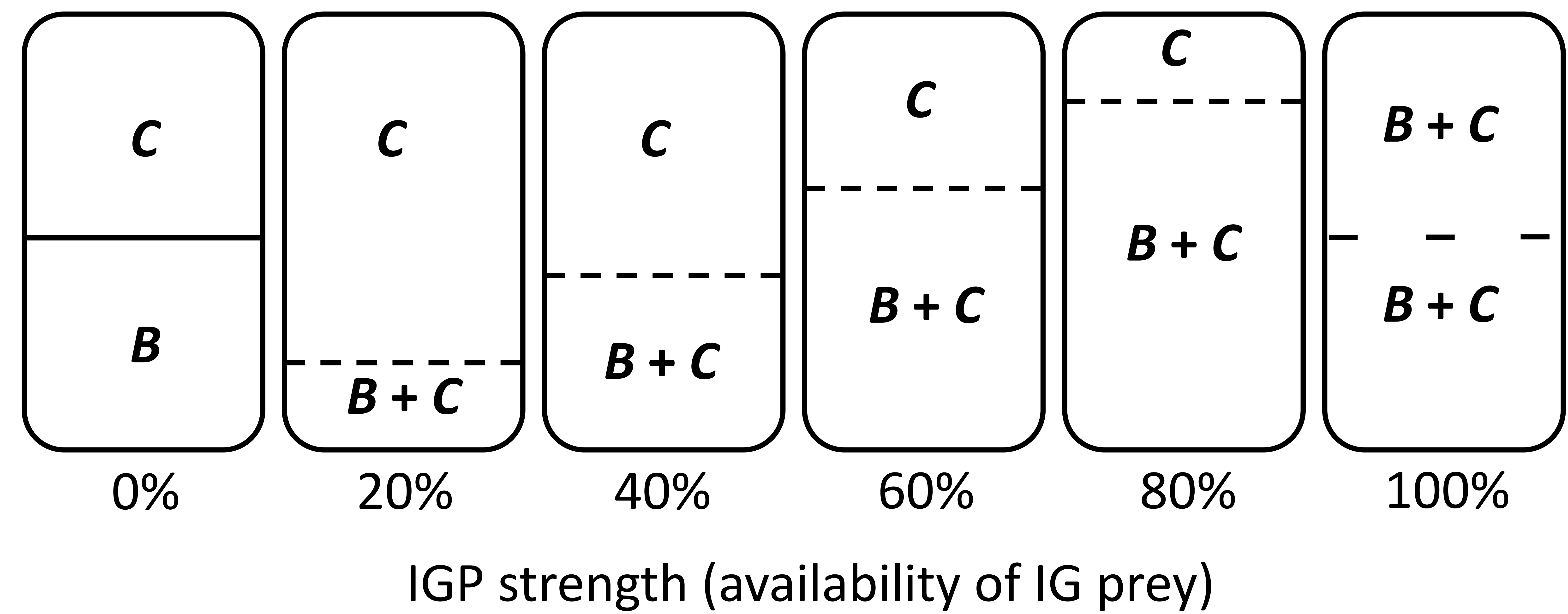
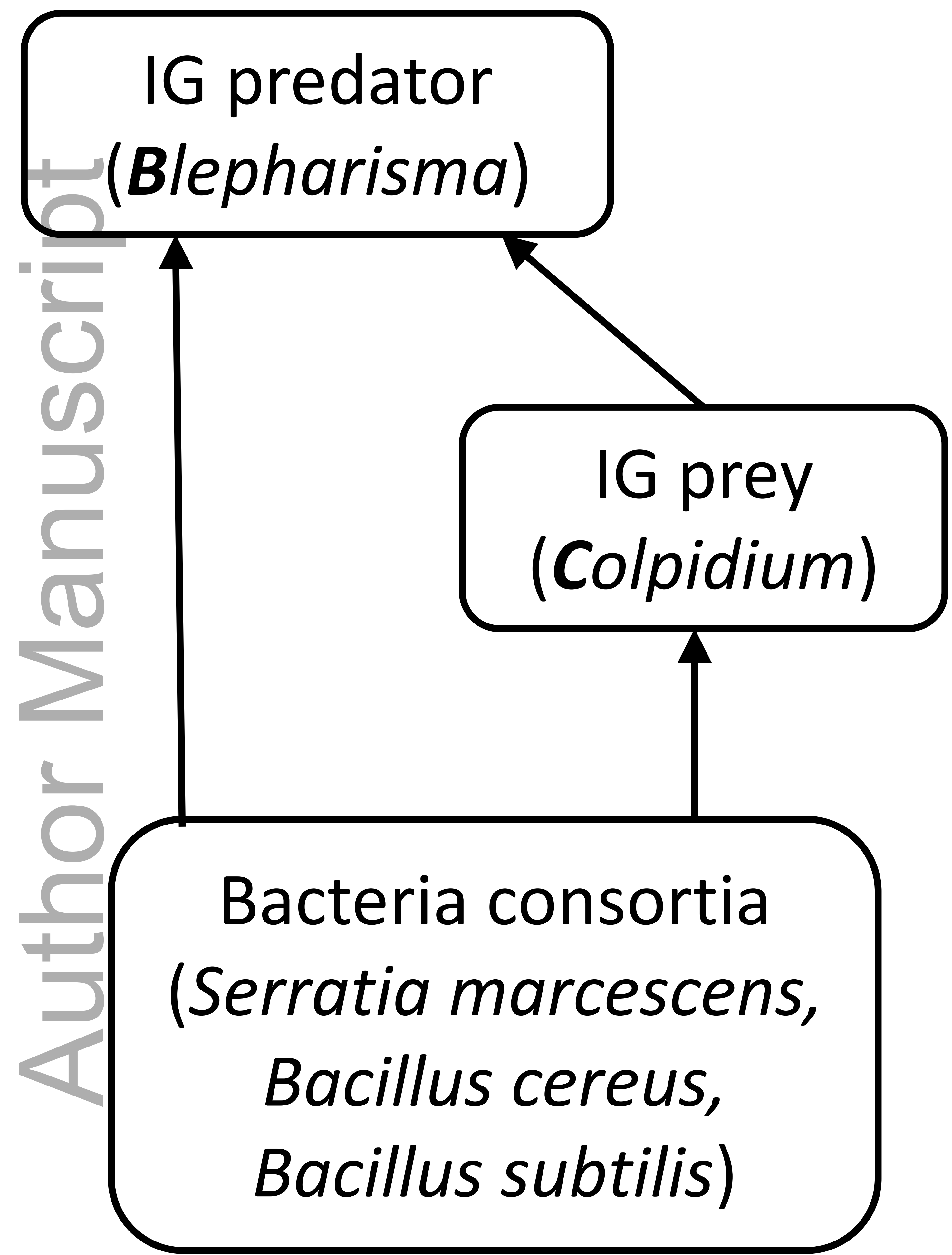


Figure 4

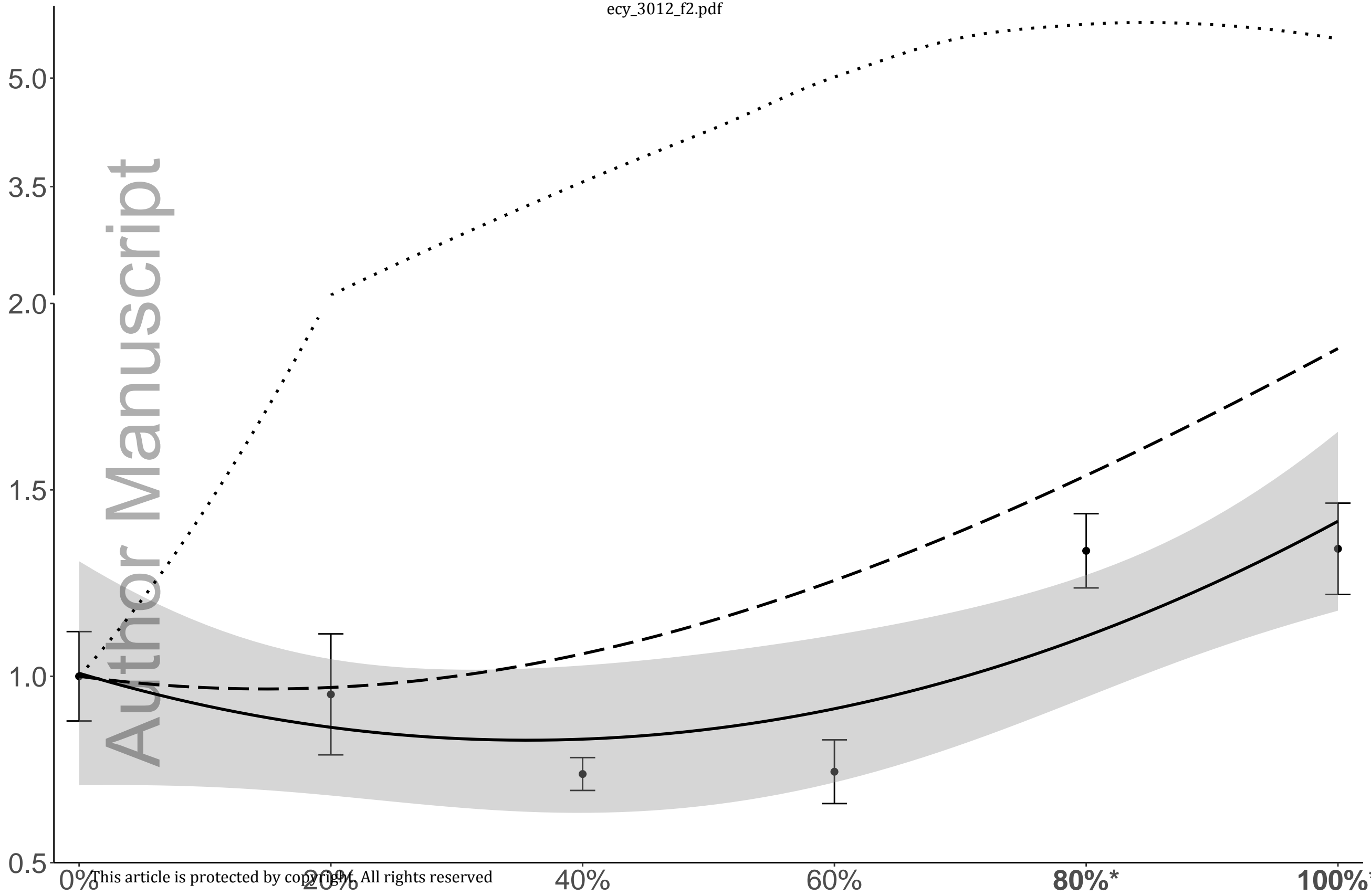




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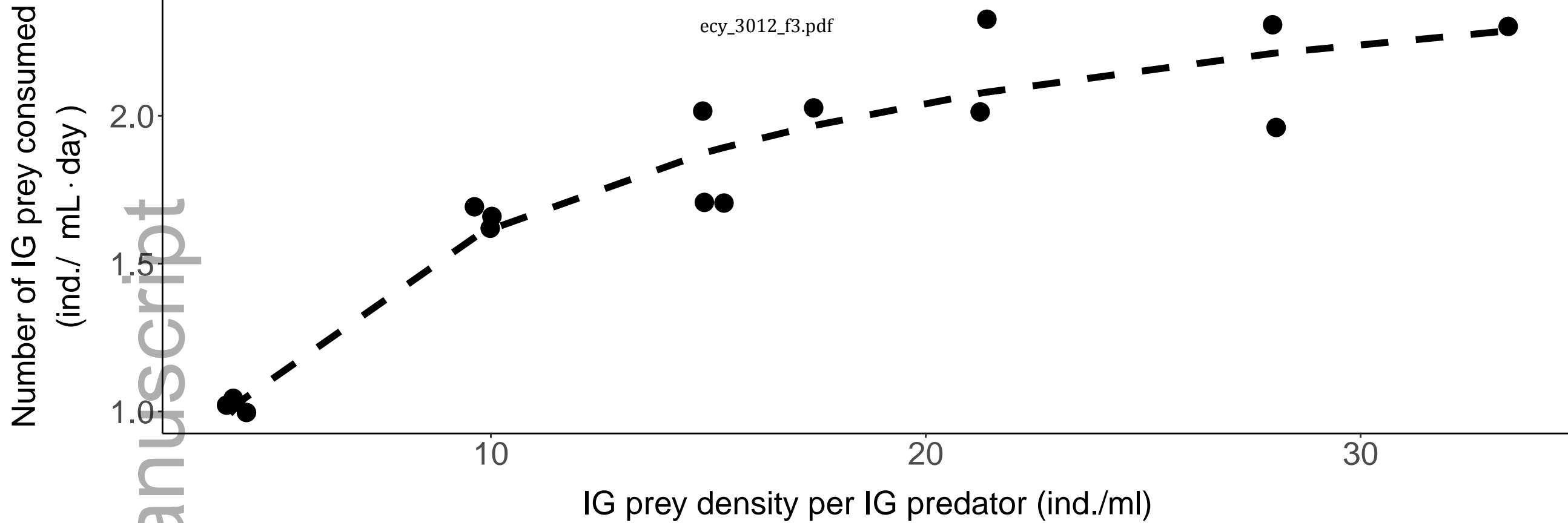
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error bars represent standard error of the mean

a.



b.

