

Experimental duration and predator satiation levels systematically affect functional response parameters

Yuanheng Li  <https://orcid.org/0000-0003-4071-8212>, Björn C. Rall  <https://orcid.org/0000-0002-3191-8389>, Gregor Kalinkat  <https://orcid.org/0000-0003-3529-5681>

DOI

<https://doi.org/10.1111/oik.04479>

Original publication date

25 October 2017 (Version of record online)

Document version

Accepted version

Published in

Oikos

Citation

Li Y, Rall BC, Kalinkat G. Experimental duration and predator satiation levels systematically affect functional response parameters. *Oikos*. 2018;127(4):590-8.

Disclaimer

This is the peer reviewed version of the following article: Li Y, Rall BC, Kalinkat G. Experimental duration and predator satiation levels systematically affect functional response parameters. *Oikos*. 2018;127(4):590-8 which has been published in final form at <https://onlinelibrary.wiley.com/doi/full/10.1111/oik.04479>. This article may be used for non-commercial purposes in accordance with Wiley Terms and Conditions for Self-Archiving.

1 Experimental duration and predator satiation levels
2 systematically affect functional response parameters

3 Yuanheng Li ^{*1,2,3}, Björn C. Rall^{†1,2}, and Gregor Kalinkat^{‡4}

4 ¹German Centre for Integrative Biodiversity Research (iDiv)

5 Halle-Jena-Leipzig, Deutscher Platz 5e, 04103 Leipzig, Germany

6 ²Institute of Ecology, Friedrich Schiller University Jena, Dornburger-Str.
7 159, 07743 Jena, Germany

8 ³Johann-Friedrich-Blumenbach Institute of Zoology and Anthropology,
9 Georg-August-Universität Göttingen, Berliner Str. 28, 37073 Göttingen,
10 Germany

11 ⁴Department of Biology and Ecology of Fishes, Leibniz-Institute of
12 Freshwater Ecology and Inland Fisheries, Müggelseedamm 310, 12587
13 Berlin, Germany

14 October 16, 2017

15

^{*}yuanheng.li@idiv.de; ORCID: <http://orcid.org/0000-0003-4071-8212>

[†]bjoern.rall@idiv.de; ORCID: <http://orcid.org/0000-0002-3191-8389>

[‡]kalinkat@igb-berlin.de; ORCID: <http://orcid.org/0000-0003-3529-5681>

Abstract

Empirical feeding studies where density-dependent consumption rates are fitted to functional response models are often used to parameterize the interaction strengths in models of population or food-web dynamics. However, the relationship between functional response parameter estimates from short-term feeding studies and real-world, long-term, trophic interaction strengths remains largely unexamined. In a critical first step to address this void, we tested for systematic effects of experimental duration and predator satiation on the estimate of functional response parameters, namely attack rate and handling time. Analyzing a large data set covering a wide range of predator taxa and body masses we show that attack rates decrease with increasing experimental duration, and that handling times of starved predators are consistently shorter than those of satiated predators. Therefore, both the experimental duration and the predator satiation level have a strong and systematic impact on the predictions of population dynamics and food-web stability. Our study highlights potential pitfalls at the intersection of empirical and theoretical applications of functional responses. We conclude our study with some practical suggestions for how these implications should be addressed in the future to improve predictive abilities and realism in models of predator-prey interactions.

Keywords: Type II functional response — handling time — attack rate — experimental duration — predator satiation level — diurnal cycle — digestive limit

36 Introduction

37 Understanding species interactions and how they shape communities and ecosystems is a
38 core topic in ecological research. Trophic interactions are fundamental for ecosystems,
39 as they determine energy flow and nutrient cycling in ecological networks (Elton, 1927;
40 Brown et al., 2004; Thompson et al., 2012). Moreover, interaction strengths play a crucial
41 role in determining population dynamics and stability of food webs (May, 1972; Oaten
42 and Murdoch, 1975; Oksanen et al., 1981; Rall et al., 2008; Brose, 2010; Kalinkat et al.,
43 2013; Li et al., 2017). Functional response models which describe per capita feeding
44 rates of consumers as a function of resource densities (Solomon, 1949; Holling, 1959b)
45 provide a widely applied and standardized way to quantify these interaction strengths in
46 food webs (Berlow et al., 2004; Kalinkat et al., 2013). Accordingly, interaction strengths
47 are typically quantified by empirical studies, carried out mostly in the laboratory, from
48 which feeding data is collected and used to fit a functional response model (Jeschke
49 et al., 2002, 2004; Rall et al., 2012). Parameters from these models can then be used to
50 parameterize the interaction strengths in theoretical food web models. Hence, functional
51 response models often serve as the connection between studies of short-term, individual-
52 level interactions and long-term, community-level studies (e.g. Kalinkat et al., 2013).
53 However, most functional response studies only investigate feeding over a short portion
54 of a species lifetime, from minutes (e.g. Schröder et al., 2016) to a few days (e.g. Buckel
55 and Stoner, 2000), and the results are often applied to studies modeling interactions
56 over many generations (e.g. hundreds of years; Fox and Murdoch, 1978). Functional
57 responses quantify the feeding rate of a predator averaging over the experimental duration,
58 which is important for predicting numerical response (how predator density changes as a
59 function of prey density over time) using, e.g. mathematical models (Okuyama, 2013).
60 The combination of functional response and numerical response results in total response,

61 the rate of overall prey population consumed by predator population (Holling, 1959a).
62 Therefore, in community level studies the feeding rates have to represent the average
63 consumption situation over the lifetime of a predator or many generations, not the feeding
64 rates from a feeding trail that only includes a feeding event when the predator is hungry.
65 Whether functional response parameter values derived from short-term functional response
66 experiments hold for longer periods (e.g. on ecological time scales relevant to population
67 dynamics) remains largely unexamined (but see Fox and Murdoch, 1978).

68 In a similar vein, the satiation levels of predators prior to feeding studies also may
69 modify functional response parameter estimates. As a predator's satiation level directly
70 affects its motivation to forage (Jeschke, 2007), satiated predators are expected to consume
71 fewer prey individuals than starved predators which in turn would alter the functional
72 response parameters. We addressed whether and how the experimental duration and the
73 satiation level of predators affects the estimates of functional response parameters using a
74 literature based functional response data base (Rall et al., 2012).

75 Due to the availability of data we focus our analysis on type II functional responses as
76 described by Holling's disc equation (Holling, 1959b). This is the most widely-applied
77 functional response model (Jeschke et al., 2002, 2004; Kalinkat and Rall, 2015), where
78 the per capita feeding rate, $f(N)$, is formulated as a function of prey density, N with
79 two parameters, instantaneous rate of searching for prey, a (hereafter: attack rate) and
80 handling time, h :

$$81 \quad f(N) = \frac{aN}{1 + ahN} . \quad (1)$$

82 In this model, Holling (1959b) assumed that a predator spends its whole time budget on
83 foraging, which includes activities such as searching, capturing, subduing, ingesting and

84 digesting the prey. The attack rate, a , describes the space (i.e. area or volume, depending
 85 on interaction type; Pawar et al., 2012; Barrios-O'Neill et al., 2016) that a predator
 86 searches per unit of time, representing the activity of searching mentioned above. The
 87 handling time, h , associated with processing the prey, describes the average time that a
 88 predator spends on a caught prey item, i.e. subduing, ingesting and digesting. These two
 89 parameters also determine the shape of the functional response curve, where the attack
 90 rate determines the feeding rate at low prey densities and the handling time determines
 91 the maximum feeding rate (Fig. 1).

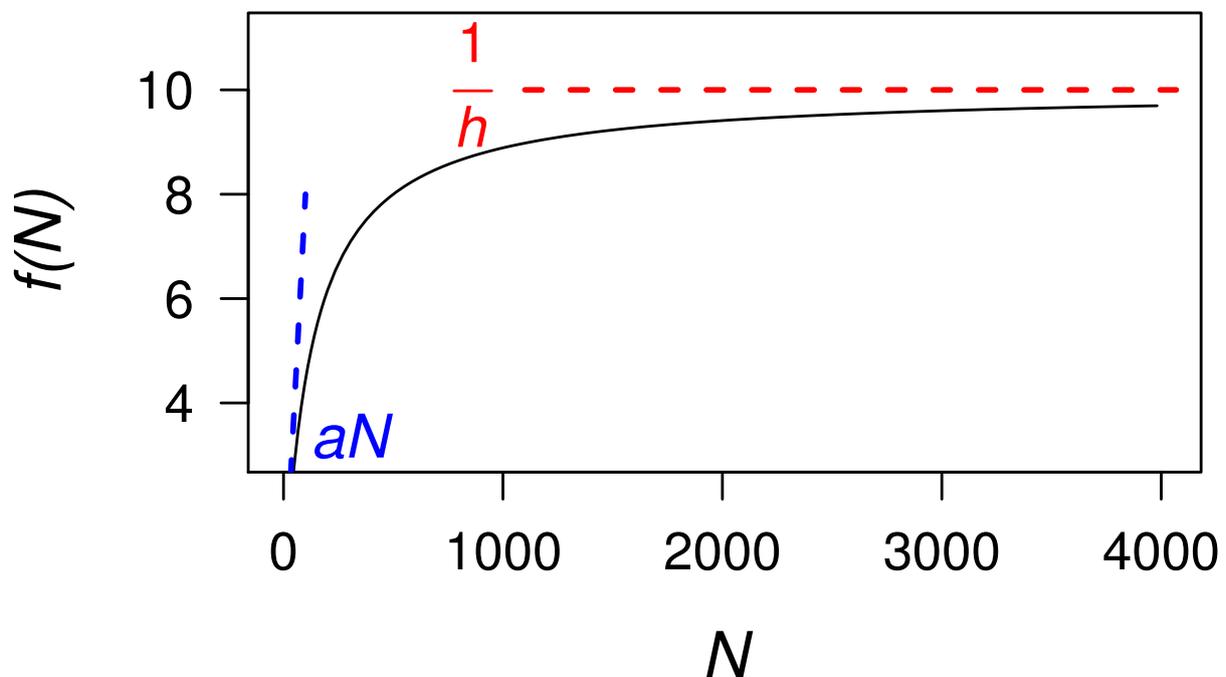


Figure 1: Schematic curve of type II functional response. The red dashed line denotes the inverse of handling time, $\frac{1}{h}$ (see Eq. (1)) which sets the limit of maximum feeding rate. The blue dashed line denotes the tangent line of the curve at the minimal prey density, N which describes the potential increase of feeding with prey density around low prey densities. This potential increase around low prey densities is determined by the attack rate, a (see Eq. (1)).

92 Attack rates and handling times indirectly derived through fitting functional response
 93 models to feeding data often do not resemble the attack rates and handling times derived
 94 by direct observation (Mols et al., 2004; Jeschke and Tollrian, 2005; Sentis et al., 2013;
 95 but see Tully et al., 2005; Hauzy et al., 2010). As there are more activities than ‘searching

96 for prey' and 'subduing the prey' in the life histories or even diurnal cycles of predators
97 (e.g. active and resting periods), a plethora of biological (i.e. physiological and behavioral)
98 processes are collapsed into the attack rate and the handling time (Jeschke et al., 2002;
99 Jeschke and Tollrian, 2005; Casas and McCauley, 2012). Even in a predator's activity
100 period it may not spend the whole time on foraging. For example, grazing ruminants feed
101 in a discrete fashion rather than continuous grazing, i.e. they switch between grazing
102 and resting (Gregorini et al., 2006). As Holling's (1959b) disc equation does not have
103 any term accounting for other activities, e.g. rest or sleeping, handling times and attack
104 rates have to incorporate those time budgets in cases where these other activities apply.
105 Parameter estimates in a long-term functional response experiment are therefore much
106 more likely to incorporate non-foraging behaviours than estimates derived from a short-
107 term experiment using the same predator-prey pair. Specifically, the feeding rates derived
108 from the long-term study would be lower than those from the short-term study (on the
109 condition that no change in life history traits, e.g. moults). Lower feeding rates will
110 likely affect functional response parameter estimation, decreasing the attack rate and
111 increasing handling time estimates (Fig. 2a and 2b). Mathematically, the feeding rate,
112 $f(N)$ is negatively related to the handling time, h (Fig. 1) and the increased handling
113 times in long-term experiments (where feeding rate should be lower) is expected for this
114 reason. As the attack rate accounts for the average successful search rate for the entire
115 experimental duration, increasing experimental duration which generally includes more
116 time for other activities than foraging, would lead to reduced attack rates (Casas and
117 McCauley, 2012).

118 Foraging motivation is also expected to be influenced by predator satiation (Jeschke,
119 2007). Essington et al. (2000) separate the effect of predator satiation to act on two
120 temporal scales: 1) instantaneous satiation occurs when feeding rate exceeds gut capacity

121 (constraint of gut size) and 2) integrated satiation occurs when feeding rate exceeds the
122 time required to digest prey (constraint of digestion rate). The longer handling times
123 associated with satiated predators may mostly reflect the constraint of digestion rate, and
124 the comparably shorter handling times of starved predators may be caused by a lack of
125 constraint from gut size. In extreme cases, a predator with a fully-filled gut will be unable
126 to feed even with infinite food supply, a well-known phenomenon called ‘digestive limit’
127 (Kleiber, 1961; Herbers, 1981). It describes a phenomenon that consumers are able to
128 fill-up their guts and meet their energy requests rather easily, e.g. on a daily basis (Jeschke,
129 2007). According to this physiological phenomenon, gut sizes and digestion rates could be
130 limiting factors for the resulting (maximum) feeding rates and associated handling times
131 in the functional response models (Jeschke et al., 2002, 2006). ‘Digestive limits’ have been
132 demonstrated in a range of vertebrate species but only few invertebrates (Karasov and
133 McWilliams, 2005; Jeschke and Tollrian, 2005; Jeschke, 2007). Under the assumption that
134 digestive limits are a rather general mechanism holding for most consumers, the satiation
135 level of a predator before a feeding study will influence the estimate of handling time
136 (Anderson et al., 1978; Jeschke et al., 2002; Jeschke, 2007). Testing pre-fed predators in
137 feeding trials would then lead to longer handling times compared to testing starved ones.
138 The time budgets of the handling time of a satiated predator would involve not only the
139 time for killing (t_{kill}) and ingesting (t_{ing}), but also the time for digestion (t_{dig}) and even
140 time for other activities (see Fig. 2c.)

141 As the experimental duration increases, the probability of reaching satiation would
142 increase systematically for every efficiently foraging predator. Thereafter, if the experi-
143 mental duration is long enough and prey is sufficient, the predator can reach satiation
144 and the constraint of digestive limit is very strong. In this case, the handling time could
145 be influenced by including the time budget for digestion. As the experimental duration

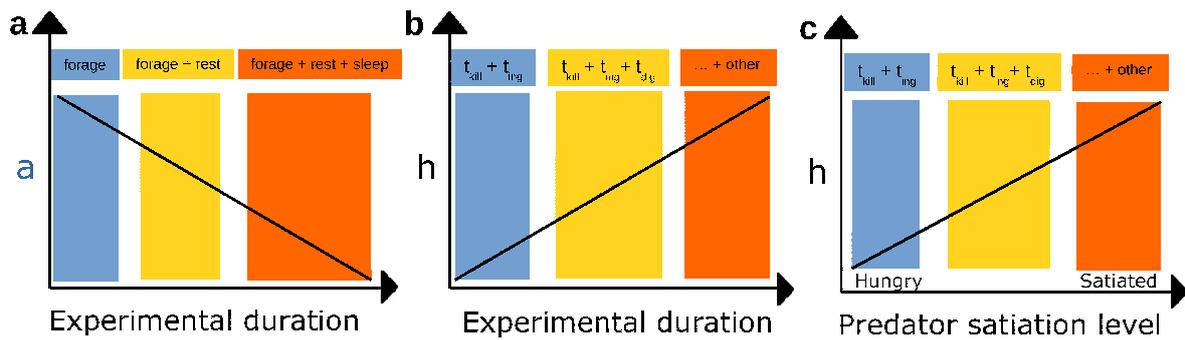


Figure 2: The potential effects of experimental duration (panel **a** and **b**) and satiation level of the predator (panel **c**) on functional response parameter estimates. We hypothesized that increasing experimental duration would lead to decreasing attack rates, a (Eq. (1), panel **a**). Elongated experiments may lead to increased handling times, h (Eq. (1), panel **b**). We also hypothesized that a satiated (pre-fed) predator shall result in longer handling times (h) than hungry (starved) predators (panel **c**). The text “other” in panels **b** and **c** denotes time spent on non-feeding activities, e.g. sleep.

146 further increases, other activities of the predator (e.g. sleep) could be involved. In this case,
 147 elongated experimental duration can additionally increase handling time by incorporating
 148 a growing proportion of non-feeding activities.

149 For this study, we used a data set from Rall and colleagues (2012) and updated it
 150 with information on starvation and experimental time. We focused on type II functional
 151 responses leaving 451 distinct data points from 61 peer-reviewed publications. The data
 152 mostly consists of controlled laboratory experiments (99%) with arthropods (78%) and
 153 vertebrates (17%) as predators. Prior to our analyses we hypothesized that, 1) experimen-
 154 tal duration has systematic effects on functional response parameters, particularly on the
 155 attack rate, and that, 2) the influence of predator satiation on handling time holds over a
 156 wide range of different taxa, body masses, and dimensionality of consumer search space.
 157 As elaborated above, we assume that in general, satiated predators should consume fewer
 158 prey than hungry ones on the premise that all other conditions are the same. Therefore,
 159 3) the handling time of satiated predators should be longer than that of hungry ones as
 160 it should incorporate additional time budgets for digestion and activities unrelated to

161 foraging.

162 **Methods**

163 **Data and statistical analysis**

164 We analyzed a data set of published functional responses from empirical studies (Rall
165 et al., 2012). To be included in our analyses studies needed to report experimental
166 duration, consumer and resource body masses, as well as experimental temperatures, as
167 these are main drivers for functional response parameter estimates (Hansen et al., 1997;
168 Rall et al., 2012; Kalinkat et al., 2013; Kalinkat and Rall, 2015). Additionally, we checked
169 and included information on the satiation levels of predators. The predator satiation is
170 represented by “feeding-or-not” prior to the studies, i.e. ‘fed’ for the predators which were
171 fed before the feeding trials and ‘starved’ for the predators which were isolated from food
172 source before the feeding trials.

173 In order to assemble the data set we excluded functional responses derived from
174 experiments which 1) lacked information on experimental duration or predator satiation
175 levels, 2) were not type-II functional responses and 3) excluded ones that are for parasitoids
176 (not suitable for testing predator satiation). The final data set consisted of 451 functional
177 responses from 61 studies (see the full bibliography in Appendix I). It spans 14 orders of
178 magnitude of predator body-mass and covers predator species from 28 taxonomic orders
179 (see Appendix I). It includes 338 and 113 functional responses for starved and fed predators,
180 respectively, and it includes data on experimental duration ranging from 0.08 h to 240 h in
181 which 67.6% are exactly 24 h. It also includes functional responses for studies performed
182 in two- and three-dimensional spaces, in which 243 were two-dimensional interactions

183 and 208 were three-dimensional interactions. We chose to also analyze dimensionality as
184 the units of attack rates are different in two- and three-dimensional spaces (i.e. $[\text{m}^2 \text{s}^{-1}]$
185 and $[\text{m}^3 \text{s}^{-1}]$) which might also cause varying scaling relationships (Pawar et al., 2012;
186 Barrios-O'Neill et al., 2016).

187 In the following steps, we analyzed the functional response parameters attack rate,
188 a [$\text{m}^2 \text{s}^{-1}$ | $\text{m}^3 \text{s}^{-1}$], and handling time, h [s] in relation to experimental duration, t_e [s]
189 and predator satiation, S (starved, S_y or fed, S_n). To account for strong effects of
190 predator body mass (M [mg]), temperature (T [K]) and dimensionality ($2D/3D$) we also
191 added these as explanatory variables (Rall et al., 2012; Pawar et al., 2012). Prior to the
192 analysis we ln-transformed the explanatory variables body mass and experimental duration
193 and applied an Arrhenius transformation to temperature ($\frac{T-T_0}{kT_0}$), where k [eV K^{-1}] is
194 the Boltzmann's constant and T_0 (293.15 K, 20 °C) is the normalization temperature.
195 Therefore, the intercepts of temperature scalings are shifted to the values at normalization
196 temperature (for more details see Gillooly et al., 2001; Rall et al., 2012). We also tested
197 the collinearity between independent variables, a test on variance inflation factor (VIF)
198 (Zuur et al., 2010). This test was operated in R (R Core Team, 2016) using the 'corvif'
199 function provided by Zuur et al. (2009). According to Zuur et al. (2010), the independent
200 variables of a linear mixed-effects model shall all have the values of VIF less than 3
201 to make sure that there are not correlation between them (i.e. multicollinearity). The
202 result showed that there is no collinearity between any independent variables, the VIF of
203 predator mass 1.64, temperature 1.25, experimental duration 1.33, predator satiation 2.19
204 and dimensionality 2.18.

205 Thereafter, we used Bayesian Information Criterion (BIC; Zuur et al., 2009, p. 121) to
206 select the optimal random structures of the linear models which were fitted according
207 to restricted maximum likelihood ('REML' Pinheiro et al., 2016, for more details see

208 Appendix II). We analyzed the data with linear mixed-effects models ('lme' function in
 209 'nlme' package in R; Pinheiro et al., 2016; R Core Team, 2016). We included all pairwise
 210 interactions of the fixed variables while selecting the optimal random structure for the
 211 models of both attack rate and handling time (these full models see Appendix II). After
 212 selecting the optimal random effects structure, the BIC values for attack rate and handling
 213 time models were computed using the 'dredge' function in the "MuMIn" package in R
 214 (Bartoń, 2016). Optimal models were then selected according to the lowest BIC value
 215 following Raftery (1995). Accordingly, ΔBIC for each second best fitting model should be
 216 at least >2 ($\Delta\text{BIC}=\text{BIC} - \min(\text{BIC})$).

217 Results

218 We first selected the appropriate models based on their ΔBIC for both, attack rate
 219 (ΔBIC for second best model=15.89) and handling time (ΔBIC =7.98). The selected
 220 model showed that attack rate can be described by predator body mass, temperature,
 221 experimental duration and dimensionality (Tab. 1):

$$\ln(a_{2D}) = \ln(a_{0_{2D}}) + b \ln(M) + E_a \frac{T - T_0}{k T T_0} + i \ln(t_e) \quad (2a)$$

$$\ln(a_{3D}) = \ln(a_{0_{3D}}) + b \ln(M) + E_a \frac{T - T_0}{k T T_0} + i \ln(t_e) \quad (2b)$$

223 Handling time can be described by predator body mass, temperature and predator satiation
 224 (Tab. 1):

$$\ln(h_{S_y}) = \ln(h_{0_{S_y}}) + c \ln(M) + E_h \frac{T - T_0}{k T T_0} \quad (3a)$$

225

$$\ln(h_{S_n}) = \ln(h_{0_{S_n}}) + c \ln(M) + E_h \frac{T - T_0}{k T T_0} \quad (3b)$$

226 In equations above (Eqs. (2), (3)), a_0 and h_0 are constants, b and c are the scaling
 227 exponents for predator body mass, M [mg], E_a and E_h [eV] are activation energies
 228 describing the exponents of temperature and i represents the scaling exponent of attack
 229 rate for experimental duration. The attack rate scaled negatively with experimental
 230 duration and its model included the influence of dimensionality on its intercepts (panel A,
 231 B and C of Fig. 3, Tab. 1). The handling time included the effect of predator satiation
 232 level, resulting in different constants for starved and fed predators, respectively (panel D
 233 and E of Fig. 3, Tab. 1).

Table 1: Statistical results for attack rate and handling time. All interaction terms have been excluded by model selection (see Methods for details).

	Variable ^a		Estimate	S.E.	p-value
attack rate	dimension	$\ln(a_{0_{2D}})$	-10.59	1.96	< 0.01
		$\ln(a_{0_{3D}})$	-12.65	1.93	< 0.01
	predator mass	b	0.49	0.08	< 0.01
	temperature	E_a	0.43	0.06	< 0.01
	experimental duration	i	-0.56	0.18	< 0.05
	predator satiation	excluded			
handling time	predator satiation	$\ln(h_{0_{S_y}})$	10.64	0.44	< 0.01
		$\ln(h_{0_{S_n}})$	13.01	0.73	< 0.01
	predator mass	c	-0.73	0.05	< 0.01
	temperature	E_h	-0.30	0.10	< 0.01
	experimental duration	excluded			

^asee Eq. (2), (3)

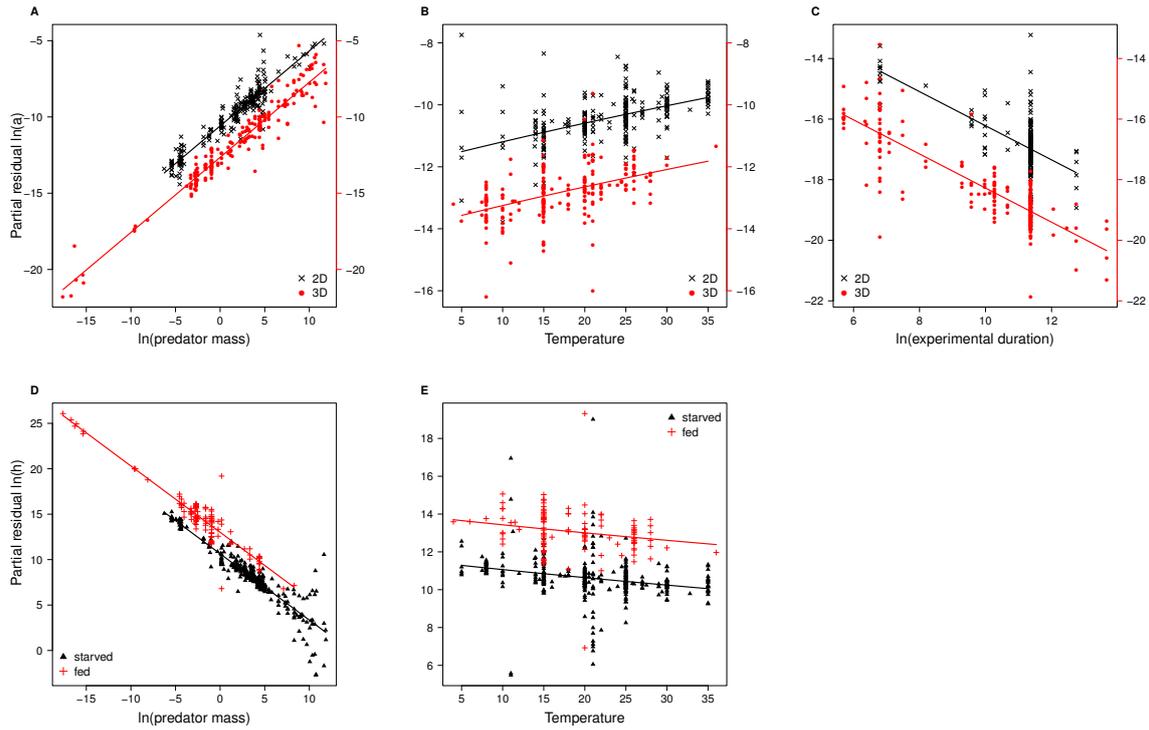


Figure 3: Dependencies of attack rate and handling time. Partial residuals are plotted on the y-axes and all variables other than temperature were ln-transformed. (For details on the derivation of partial residuals see Appendix II). The attack rate (a , [$\text{m}^2 \text{s}^{-1} \mid \text{m}^3 \text{s}^{-1}$]) increases with predator body mass [mg] (panel A), temperature [$^{\circ}\text{C}$] (panel B), and decreases with experimental duration [s] (panel C; see Tab. 1). Data of attack rates in two- and three-dimensional cases are plotted with different colors as they have different units (see legends). Handling time (h) decreases with predator body mass (panels D) and temperature (panels E), while handling times for fed predators are longer than those for starved predators (see legends and Tab. 1).

Discussion

Here we used a large data set of empirical functional responses to investigate if the experimental duration and satiation level of predators have systematic effects on the estimates of functional response parameters. We included studies where feeding data were fitted to the most widespread ‘type II functional response’ model. Notably, the resulting data set contains predator-prey pairs from a wide variety of ecosystem types including the marine, freshwater and terrestrial realm, as well as a wide range of taxa, from protists to vertebrates. Our results demonstrate that attack rate estimates decrease with increasing experimental duration and that the handling times of satiated predators are longer than those of hungry predators. Thus, two of our hypotheses were supported by our findings (Fig. 2a, c), whereas our hypothesis that increasing experimental duration increases handling time was not supported (Fig. 2b).

Our analyses of attack rates illustrated the influence of predator body mass, temperature and experimental duration. While the results of the effects of predator body mass and temperature on attack rates are consistent with previous studies (Rall et al., 2012; Fussmann et al., 2014) we show here that there is a general effect of experimental duration on the estimates of attack rates that holds across a wide range of taxa and body masses. The finding that attack rate decreases with experimental duration is intuitive to understand and can be attributed to biological mechanisms (Jeschke et al., 2002; Jeschke, 2007). In general, it suggests that the shorter the experimental duration relative to the generation time of the predators, the greater the potential for laboratory experiments to miss important time constraints on foraging. Longer experimental durations will automatically involve a higher proportion of non-feeding activities in foraging experiments. Within a diurnal cycle (24 hour period), the majority of ‘other activities’ consists of resting and sleeping for most animals (Campbell and Tobler, 1984). Therefore, assuming

259 all other conditions are kept constant (e.g. identical predator-prey pair with constant
260 size ratios, identical and standardized satiation levels of the predators), the attack rate
261 estimates derived from a feeding study of 24 hours will be smaller than those obtained
262 through a short-term experiment that includes only the high-activity window out of the
263 diurnal cycle of a given predator (Casas and McCauley, 2012). Despite the suggestion that
264 gut sizes of some predators can be somewhat phenotypically plastic (over periods of weeks
265 or months; Karasov and McWilliams, 2005; Van Gils et al., 2005), we still found that the
266 handling times for satiated predators are longer than those for starved predators. If the
267 gut plasticity plays a big role, we shall have found the opposite, i.e. the starved predators
268 would have had reduced gut capacities and thus showed longer handling times. 96 % of
269 the data used in this study are from experiments within 24 hours. Therefore our findings
270 on the effects of experimental duration are not confounded by changes in gut capacity
271 (Fig. 2a). In addition, if a functional response experiment lasts for the whole generation
272 time of the predator, the resulting functional response parameters are supposed to be a
273 good match to those estimated from long-term population dynamics, e.g. experiments
274 with protists (DeLong et al., 2014).

275 There are a few empirical case studies also supporting our findings relating to the
276 effects of experimental duration. For instance, Fox and Murdoch (1978) tested how
277 functional responses of a predatory water bugs (*Notonecta hoffmanni*) vary between
278 short-term (3 hours) and long-term (12 hours) experiments (Fox and Murdoch, 1978).
279 Even though Fox and Murdoch (1978) did not perform a statistical analysis to compare the
280 estimates of functional response parameters between short- and long-term experiments, the
281 estimated values for attack rates are consistent with our results. Another recent modeling
282 study confirmed this effect of experimental duration on the estimates of attack rates and
283 explicitly highlighted that the inclusion of different activities during diurnal cycles may

284 bias attack rate estimation (Casas and McCauley, 2012). Our results indicate that these
285 findings of Casas and McCauley (2012) are likely generalizable to most predator-prey pairs.
286 For future studies, it would also be important to address how longer feeding trials, over
287 several weeks or even months, will affect the estimation of interaction strengths (Buckel
288 and Stoner, 2000).

289 Our results did not show an effect of predator satiation level on attack rates. Jeschke
290 (2007) mentioned that one way to affect the attack rate is through affecting foraging
291 motivation of the predators (Barrios-O'Neill et al., 2016). The foraging motivation of the
292 predators, in turn, can be affected by stressors or distractors on predators, e.g. threat from
293 a top-predator. We can only speculate that in the environments of most functional response
294 experiments, there are no such external stressors on predators to affect their foraging
295 motivations (but see Toscano and Griffen, 2014; Wasserman et al., 2016). Therefore,
296 we might expect significant effects of predator satiation on foraging motivation only if
297 external stressors like top-predators enhance the differences.

298 Our statistical results documented systematic influence of predator body mass, tem-
299 perature and predator satiation levels on handling times. Notably, experimental duration
300 had no effect on handling times. With a data set that includes both invertebrate and
301 vertebrate predators, we showed that the estimates of handling times for starved predators
302 were lower than those for the fed ones. Previous studies suggested the influence of satiation
303 level on handling times mostly for vertebrate predators (Karasov and McWilliams, 2005;
304 Jeschke and Tollrian, 2005; Jeschke, 2007). Particularly, Anderson et al. (1978) is one
305 of few experimental studies which explicitly tested how predator satiation level affects
306 the functional response. There, the authors demonstrated that zebra fish (*Danio rerio*)
307 showed considerably higher maximum feeding rates when they were starved for 24 hours
308 before the experiment compared to satiated fish fed one hour before the trial (Anderson

309 et al., 1978). Here we generalized this finding to invertebrates, as the majority 78 % of the
310 data we analyzed are from arthropod predators. This supports the theoretical assumption
311 that generally, both vertebrate and invertebrate predators may face digestion limits (see
312 also Jeschke et al., 2002; Jeschke, 2007). In one of the rare experimental studies addressing
313 this issue for invertebrates, however, Maselou et al. (2015) found for a predatory mirid bug
314 (*Macrolophus pygmaeus*) that the estimates of functional response parameters were not
315 affected by predator satiation. This might be due the specific design where four different
316 treatments of gradually differing starvation levels were tested, while a treatment including
317 fully satiated predators was missing. Moreover, all four functional response curves in
318 this study did not seem to reach full satiation (Maselou et al., 2015). The comparison
319 between satiated and starved predators seems to be important for addressing the effects
320 of predator satiation level on functional response parameter estimates. Another study
321 investigated the influence of predator satiation with data of predatory fish (largemouth
322 bass, *Micropterus salmoides*; Essington et al., 2000). In agreement with our finding, the
323 authors state that feeding rates are reduced by predator satiation (Essington et al., 2000).

324 To better address the effect of predator satiation on functional response parameters
325 in future studies, we shall keep in mind the effect of predator satiation acting on two
326 different temporal scales: Essington et al. (2000) suggests separating the effect of predator
327 satiation to act on two temporal scales: 1) instantaneous satiation which reflects the
328 constraint of gut size and 2) integrated satiation which reflects the constraint of digestion
329 rate (Essington et al., 2000) which is in line with suggestions by Jeschke and colleagues
330 (2002; 2006). The longer handling times associated with satiated predators may mostly
331 reflect the constraint of digestion rate, and the comparably shorter handling times of
332 starved predators may be caused by a lack of constraint from gut size. We used binary
333 data of predator satiation rather than the absolute time that the predator was starved

334 prior of the experiments due to two reasons: 1) for some data that we had in this dataset,
335 the absolute starvation time are lacking; 2) an alternative analysis with a restricted dataset
336 including only data points where absolute starvation time was available showed similar
337 results than the analysis of the complete dataset with binary data (alternative analysis
338 not shown here). In future studies, a predator's starvation time that is relative to predator
339 body mass or a starvation time that is relative to predator's life span is worthy to be
340 analyzed. These specific starvation time may help to spread the cluster of data points
341 where experimental duration is around 24 hours and give better insight to the question.

342 Our meta-analysis showed that attack rates of two- and three-dimensional experiments
343 demonstrate the same slope using the ln-transformed linear model. Thus Pawar et al.
344 (2012) found different slopes of attack rates for two- and three-dimensional cases. One
345 obvious difference is that their dataset included not only type II but also type III
346 functional responses. Therefore, Pawar et al. (2012) showed that the attack rates at
347 minimum measured prey densities. This could be one reason why we obtained a different
348 result. Type III functional responses might respond differently to dimensionality than
349 type II functional responses, for instance Type III functional response could indicate a
350 complex spatial structure. Therefore, different dimensionalities together with the complex
351 transition between type II and type III responses might cause these varying responses
352 of attack rates (Barrios-O'Neill et al., 2016; Li et al., 2017). In contrast, our dataset is
353 dominated by data from laboratory functional response experiments where unstructured,
354 artificial habitat prevails (Rall et al., 2012; Kalinkat and Rall, 2015).

355 Empirical studies that aim to quantify interaction strengths are time-consuming
356 and often need extensive replication to investigate how particular effects drive attack
357 rates, handling times, and other parameters in more complex functional response models
358 (Kalinkat et al., 2013; Barrios-O'Neill et al., 2016). Achieving high replication of long-term

359 experiments that are close to natural conditions will most often be logistically infeasible.
360 To that end our study demonstrates that short-term functional response studies will most
361 likely lead to overestimated interaction strengths in models of predator-prey dynamics
362 or food webs. However, our results also demonstrate that this bias can be explained
363 by plausible biological mechanisms. Understanding these mechanism and incorporating
364 them when scaling up from local, short-term, studies to population, community or
365 even ecosystem-level effects holds much promise for a better understanding how species
366 interactions shape communities and ecosystems.

367 **Conclusion**

368 In the present study, we addressed the systematic effects of two common issues in feeding
369 studies, i.e. how experimental duration and satiation levels of predators affect the
370 parameter estimates in widely applied functional response models. Our study indicates
371 clear and intuitive biological mechanisms affecting the functional response parameters.
372 When models parameterized accordingly are scaled up, these effects will likely modify
373 the estimates of the dynamics and stability of populations, food webs, ecosystems, and,
374 ultimately, biodiversity. Theoretically, both higher attack rates and shorter handling times
375 will strengthen the feeding interactions in population and food-web models. Increasing
376 interaction strengths will generally lead to stronger top-down pressure where stronger
377 predator-prey interactions drive food webs into unstable conditions (Rall et al., 2008).
378 Moreover, for predator-prey systems characterized by cycling dynamics, such strengthening
379 will lead to collapse and the extinction of predator species (Rip and McCann, 2011). This
380 has important implications when realistic predictions to be applied on food-web dynamics
381 are sought. Hence both, empiricists who conduct feeding studies to estimate functional

382 response parameters, and theoreticians who try to analyze the dynamics and stability of
383 food webs often parameterized with such empirically-derived parameters should critically
384 take into account these effects. Eventually, this will enable more realistic predictions of
385 population and food-web dynamics which are crucial for understanding the consequences
386 of biodiversity loss (Brose et al., 2017) and will help to bridge lingering gaps between
387 theoretical and empirical ecological research (Jeltsch et al., 2013)

388 **Acknowledgements**

389 We are grateful to Jonathan Jeschke, Christopher Monk, Daniel Barrios-O'Neill and one
390 anonymous reviewer for their valuable comments and suggestions that helped to improve
391 this manuscript. We further thank Vicky Tröger for her assistance in revising and updating
392 the data set.

393 **Funding**

394 Funding to Y.L. was provided by the German Science Foundation (DFG) via the research
395 group "FOR 1748 - Networks on Networks". Y.L. and B.C.R. gratefully acknowledge
396 support by the German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-
397 Leipzig funded by the DFG (FZT 118). G.K. was also supported by DFG (KA 3029/2-1).
398 The funding sources had no role in study design, data collection and analysis, decision to
399 publish, or preparation of the manuscript.

400 **Conflict of interest statement**

401 The authors declare no conflict of interest.

402 **Author contribution statement**

403 G.K., B.C.R. and Y.L. designed the study; Y.L. analyzed the data in the light of discussion
404 with B.C.R. and G.K.; Y.L. wrote the manuscript with substantial further contributions
405 by B.C.R. and G.K.

406 **Data accessibility statement**

407 All data will be made available as supporting information should the manuscript be
408 accepted. R-Code for analysis will be available from the authors on request.

409 **References**

- 410 Anderson, R. M., Whitfield, P. J., Dobson, A. P. and Keymer, A. E. 1978. Concomitant
411 predation and infection processes: an experimental study. – *Journal of Animal Ecology*
412 47(3): 891–911.
- 413 Barrios-O’Neill, D., Kelly, R., Dick, J. T. A., Ricciardi, A., MacIsaac, H. J. and Emmerson,
414 M. C. 2016. On the context-dependent scaling of consumer feeding rates. – *Ecology*
415 Letters 19(6): 668–678.
- 416 Bartoń, K. 2016. MuMIn: Multi-Model Inference. – R package version 1.15.6.

- 417 Berlow, E. L., Neutel, A.-M., Cohen, J. E., de Ruiter, P. C., Ebenman, B., Emmerson, M.,
418 Fox, J. W., Jansen, V. A. A., Iwan Jones, J., Kokkoris, G. D., Logofet, D. O., McKane,
419 A. J., Montoya, J. M. and Petchey, O. 2004. Interaction strengths in food webs: issues
420 and opportunities. – *Journal of Animal Ecology* 73(3): 585–598.
- 421 Brose, U. 2010. Body-mass constraints on foraging behaviour determine population and
422 food-web dynamics. – *Functional Ecology* 24(1): 28–34.
- 423 Brose, U., Blanchard, J. L., Eklöf, A., Galiana, N., Hartvig, M., R. Hirt, M., Kalinkat,
424 G., Nordström, M. C., O’Gorman, E. J., Rall, B. C., Schneider, F. D., Thébault, E.
425 and Jacob, U. 2017. Predicting the consequences of species loss using size-structured
426 biodiversity approaches. – *Biological Reviews* 92(2): 684–697.
- 427 Brown, J. H., Gillooly, J. F., Allen, A. P., Savage, V. M. and West, G. B. 2004. Toward a
428 metabolic theory of ecology. – *Ecology* 85(7): 1771–1789.
- 429 Buckel, J. A. and Stoner, A. W. 2000. Functional response and switching behavior of
430 young-of-the-year piscivorous bluefish. – *Journal of Experimental Marine Biology and*
431 *Ecology* 245(1): 25–41.
- 432 Campbell, S. S. and Tobler, I. 1984. Animal sleep: A review of sleep duration across
433 phylogeny. – *Neuroscience & Biobehavioral Reviews* 8(3): 269–300.
- 434 Casas, J. and McCauley, E. 2012. Daily foraging cycles create overlapping time-scales in
435 functional responses. – *Oikos* 121(12): 1966–1976.
- 436 DeLong, J. P., Hanley, T. C. and Vasseur, D. A. 2014. Predator-prey dynamics and the
437 plasticity of predator body size. – *Functional Ecology* 28(2): 487–493.
- 438 Elton, C. S. 1927. *Animal Ecology*. – New York, Macmillan Co.

- 439 Essington, T. E., Hodgson, J. R. and Kitchell, J. F. 2000. Role of satiation in the functional
440 response of a piscivore, largemouth bass (*Micropterus salmoides*). – *Canadian Journal*
441 *of Fisheries and Aquatic Sciences* 57(3): 548–556.
- 442 Fox, L. R. and Murdoch, W. W. 1978. Effects of feeding history on Short-Term and
443 Long-Term Functional Responses in *Notonecta hoffmanni*. – *Journal of Animal Ecology*
444 47(3): 945–959.
- 445 Fussmann, K. E., Schwarzmüller, F., Brose, U., Jousset, A. and Rall, B. C. 2014. Ecological
446 stability in response to warming. – *Nature Climate Change* 4(3): 206–210.
- 447 Gillooly, J. F., Brown, J. H., West, G. B., Savage, V. M. and Charnov, E. L. 2001. Effects
448 of Size and Temperature on Metabolic Rate. – *Science* 293(5538): 2248–2251.
- 449 Gregorini, P., Tamminga, S. and Gunter, S. A. 2006. Behavior and Daily Grazing Patterns
450 of Cattle. – *The Professional Animal Scientist* 22(3): 201–209.
- 451 Hansen, P. J., Bjørnsen, P. K. and Hansen, B. W. 1997. Zooplankton grazing and growth:
452 Scaling within the 2-2,000- μm body size range. – *Limnology and Oceanography* 42(4):
453 687–704.
- 454 Hauzy, C., Tully, T., Spataro, T., Paul, G. and Arditi, R. 2010. Spatial heterogeneity
455 and functional response: an experiment in microcosms with varying obstacle densities. –
456 *Oecologia* 163(3): 625–636.
- 457 Herbers, J. M. 1981. Time Resources and Laziness in Animals. – *Oecologia* 49(2): 252–262.
- 458 Holling, C. S. 1959a. The components of predation as revealed by a study of small-mammal
459 predation of the European pine sawfly. – *The Canadian Entomologist* 91(5): 293–320.
- 460 — 1959b. Some characteristics of simple types of predation and parasitism. – *The Canadian*
461 *Entomologist* 91(7): 385–398.

- 462 Jeltsch, F., Blaum, N., Brose, U., Chipperfield, J. D., Clough, Y., Farwig, N., Geissler,
463 K., Graham, C. H., Grimm, V., Hickler, T., Huth, A., May, F., Meyer, K. M., Pagel,
464 J., Reineking, B., Rillig, M. C., Shea, K., Schurr, F. M., Schröder, B., Tielbörger, K.,
465 Weiss, L., Wiegand, K., Wiegand, T., Wirth, C. and Zurell, D. 2013. How can we bring
466 together empiricists and modellers in functional biodiversity research?. – Basic and
467 Applied Ecology 14(2): 93–101.
- 468 Jeschke, J. M. 2007. When carnivores are full and lazy. – *Oecologia* 152(2): 357–364.
- 469 Jeschke, J. M., Kopp, M. and Tollrian, R. 2002. Predator functional responses: dis-
470 criminating between handling and digesting prey. – *Ecological Monographs* 72(1):
471 95–112.
- 472 — 2004. Consumer-food systems: why type I functional responses are exclusive to filter
473 feeders. – *Biological Reviews* 79(2): 337–349.
- 474 — 2006. Time and energy constraints: reply to Nolet and Klaassen (2005). – *Oikos* 114(3):
475 553–554.
- 476 Jeschke, J. M. and Tollrian, R. 2005. Predicting Herbivore Feeding Times. – *Ethology*
477 111(2): 187–206.
- 478 Kalinkat, G. and Rall, B. C. 2015. Effects of climate change on the interactions between
479 insect pests and their natural enemies. – In: Björkman, C. and Niemelä, P. (eds.),
480 Climate Change and Insect Pests. CABI, pp. 74–91.
- 481 Kalinkat, G., Schneider, F. D., Digel, C., Guill, C., Rall, B. C. and Brose, U. 2013.
482 Body masses, functional responses and predator-prey stability. – *Ecology Letters* 16(9):
483 1126–1134.

- 484 Karasov, W. H. and McWilliams, S. R. 2005. Digestive constraints in mammalian and
485 avian ecology. – In: Stark, J. M. and Wang, T. (eds.), *Physiological and ecological*
486 *adaptations to feeding in vertebrates*. Science Publishers, pp. 88–112.
- 487 Kleiber, M. 1961. *The Fire of Life: An Introduction to Animal Energetics..* – Wiley.
- 488 Li, Y., Brose, U., Meyer, K. and Rall, B. C. 2017. How patch size and refuge availability
489 change interaction strength and population dynamics: a combined individual- and
490 population-based modeling experiment. – *PeerJ* 5: e2993.
- 491 Maselou, D., Perdikis, D. and Fantinou, A. 2015. Effect of hunger level on prey consumption
492 and functional response of the predator *Macrolophus pygmaeus*. – *Bulletin of Insectology*
493 68(2): 211–218.
- 494 May, R. M. 1972. Will a large complex system be stable?. – *Nature* 238(5364): 413–414.
- 495 Mols, C. M. M., Oers, K. v., Witjes, L. M. A., Lessells, C. M., Drent, P. J. and Visser, M. E.
496 2004. Central assumptions of predator-prey models fail in a semi-natural experimental
497 system. – *Proceedings of the Royal Society of London B: Biological Sciences* 271(Suppl
498 3): S85–S87.
- 499 Oaten, A. and Murdoch, W. W. 1975. Functional Response and Stability in Predator-Prey
500 Systems. – *The American Naturalist* 109(967): 289–298.
- 501 Oksanen, L., Fretwell, S. D., Arruda, J. and Niemela, P. 1981. Exploitation Ecosystems
502 in Gradients of Primary Productivity. – *The American Naturalist* 118(2): 240–261.
- 503 Okuyama, T. 2013. Consequences of variation in foraging success among predators on
504 numerical response. – *Ecology and Evolution* 3(11): 4039–4043.
- 505 Pawar, S., Dell, A. I. and Savage, V. M. 2012. Dimensionality of consumer search space
506 drives trophic interaction strengths. – *Nature* 486(7404): 485–489.

- 507 Pinheiro, J., Bates, D., DebRoy, S., Sarkar, D. and R Core Team 2016. nlme: Linear and
508 Nonlinear Mixed Effects Models. – R package version 3.1-128.
- 509 R Core Team 2016. R: A Language and Environment for Statistical Computing. – R
510 Foundation for Statistical Computing, Vienna, Austria.
- 511 Raftery, A. E. 1995. Bayesian Model Selection in Social Research. – Sociological Method-
512 ology 25: 111–163.
- 513 Rall, B. C., Brose, U., Hartvig, M., Kalinkat, G., Schwarzmuller, F., Vucic-Pestic, O. and
514 Petchey, O. L. 2012. Universal temperature and body-mass scaling of feeding rates. –
515 Philosophical Transactions of the Royal Society B: Biological Sciences 367(1605): 2923–
516 2934.
- 517 Rall, B. C., Guill, C. and Brose, U. 2008. Food-web connectance and predator interference
518 dampen the paradox of enrichment. – Oikos 117(2): 202–213.
- 519 Rip, J. M. K. and McCann, K. S. 2011. Cross-ecosystem differences in stability and
520 the principle of energy flux: Cross-ecosystem differences in stability. – Ecology Letters
521 14(8): 733–740.
- 522 Schröder, A., Kalinkat, G. and Arlinghaus, R. 2016. Individual variation in functional
523 response parameters is explained by body size but not by behavioural types in a poeciliid
524 fish. – Oecologia 182(4): 1129–1140.
- 525 Sentis, A., Hemptinne, J.-L. and Brodeur, J. 2013. Parsing handling time into its
526 components: implications for responses to a temperature gradient. – Ecology 94(8):
527 1675–1680.
- 528 Solomon, M. E. 1949. The Natural Control of Animal Populations. – Journal of Animal
529 Ecology 18: 1–35.

- 530 Thompson, R. M., Brose, U., Dunne, J. A., Hall, R. O., Hladyz, S., Kitching, R. L.,
531 Martinez, N. D., Rantala, H., Romanuk, T. N., Stouffer, D. B. and Tylianakis, J. M.
532 2012. Food webs: reconciling the structure and function of biodiversity. – *Trends in*
533 *Ecology & Evolution* 27(12): 689–697.
- 534 Toscano, B. J. and Griffen, B. D. 2014. Trait-mediated functional responses: predator
535 behavioural type mediates prey consumption. – *Journal of Animal Ecology* 83(6):
536 1469–1477.
- 537 Tully, T., Cassey, P. and Ferriere, R. 2005. Functional response: rigorous estimation and
538 sensitivity to genetic variation in prey. – *Oikos* 111(3): 479–487.
- 539 Van Gils, J. A., Dekinga, A., Spaans, B., Vahl, W. K. and Piersma, T. 2005. Digestive
540 bottleneck affects foraging decisions in red knots *Calidris canutus*. II. patch choice and
541 length of working day. – *Journal of Animal Ecology* 74(1): 120–130.
- 542 Wasserman, R. J., Alexander, M. E., Dalu, T., Ellender, B. R., Kaiser, H. and Weyl, O.
543 L. F. 2016. Using functional responses to quantify interaction effects among predators. –
544 *Functional Ecology* 30(12): 1988–1998.
- 545 Zuur, A. F., Ieno, E. N. and Elphick, C. S. 2010. A protocol for data exploration to avoid
546 common statistical problems: Data exploration. – *Methods in Ecology and Evolution*
547 1(1): 3–14.
- 548 Zuur, A. F., Ieno, E. N., Walker, N., Saveliev, A. A. and Smith, G. M. 2009. *Mixed*
549 *Effects Models and Extensions in Ecology with R*. – *Statistics for Biology and Health*.
550 Springer New York.