

Supplemental Information

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1 ODD protocol for feeding simulation model

1.1 Overview

We modeled allometric predator-prey feeding using an individual-based approach. The model description followed the ODD (Overview, Design concepts, Details) protocol (Grimm et al., 2006, 2010). The model was implemented in C++ using Code::Blocks as development environment.

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18 **1.2 Purpose**

19 The purpose of the model is to estimate the dependencies of functional response
20 parameters on patch size and habitat complexity in a system with one predator and several
21 prey items. As a preparation, we first investigated the maximum feeding rate without any
22 explicit space properties, because we assumed that the maximum feeding rate is driven by
23 physiological (mechanical) parameters such as chewing and digesting and does not scale
24 with patch size or habitat complexity. Secondly, we assessed functional responses of
25 predators to different prey densities in patches of different sizes and habitat complexity.

26 **1.3 Entities, state variables and scales**

27 One predator and several prey individuals make up the agents of the model. The common
28 state variables of predator and prey are individual identity, spatial coordinates (in cm),
29 body mass (in mg), and body mass-dependent velocity. The predator is characterized by
30 further state variables related to hunting and digestion. The parameter ‘gut-fill’ captures
31 how many milligrams of food are currently in the gut of the predator, ‘still-handling’
32 captures how many time steps are still needed for handling prey, and ‘prey-eaten’ counts
33 the number of prey items consumed by the predator.

34 The environment consists of a two-dimensional square area whose size and habitat
35 complexity can be modified. To manipulate habitat complexity in form of refuges for the
36 prey, each cell in this area is characterized by the boolean state variable ‘prey hiding’. This
37 variable is set to 1 if prey individuals staying in this cell cannot be found by a predator,
38 and to 0 otherwise. The predator and prey can move continuously in the area via random
39 walk. Predator and prey are both able to enter all the cells, but the predator is not able to

40 find prey in refuge cells. We implemented non-periodic wall-like boundary conditions
41 (Attard, 2006) to simulate a finite-sized patch.

42 In terms of scales, one grid cell always represents 1 cm * 1 cm and the spatial extent of the
43 patch can be modified from 20 cm * 20 cm to 1000 cm * 1000 cm to account for different
44 patch sizes. One time step represents one second and simulations were run for 3600 time
45 steps corresponding to one hour.

46 **1.4 Process overview and scheduling**

47 The first process that is applied in this discrete-time model is prey movement (random
48 walk with randomly chosen direction and allometric distance, i.e. the velocity of an
49 individual scales with its body mass). The following processes are the predator's decisions
50 and actions (Fig. 1 from main text). First, the predator digests prey if it has caught prey
51 and is still handling it. Subsequently, if the predator is not handling prey anymore and its
52 gut is full ('gut-fill' $\geq 60\%$), it rests, i.e. it does not take any actions. If the predator is
53 not handling prey and is hungry ('gut-fill' $< 60\%$), the predator moves according to the
54 same rules as prey. After reaching the new position, the predator investigates if it
55 encounters a prey in the cell. If there is a prey individual in the same cell and it is not
56 hidden by habitat complexity, the prey will be attacked. If the attack is successful, another
57 prey item is placed randomly into the grid to keep prey density constant. The predator
58 starts to handle the prey in the next time step.

59 1.5 Design concepts

60 *Basic principles* - This model mimics classical functional response experiments, but can
61 explore much larger patch sizes in its virtual arena than in a real laboratory. Moreover, the
62 model can more easily be used to determine maximum feeding rates than lab experiments,
63 because the amount of prey can be held constantly until maximum feeding rates are
64 reached. *Emergence* - Functional responses are the main emerging pattern from the model,
65 arising from the predator's efficiency at catching prey in patches of different sizes and
66 habitat complexity. *Sensing* - Predator and prey are able to detect each other when they
67 meet in the same cell and the cell is not a refuge cell. Both predator and prey are able to
68 detect patch edges and stop nearby when they reach an patch edge. Next time when this
69 individual needs to move, it just moves according to the same rules as before. *Interaction* -
70 Predators interact with prey by feeding on the prey when they meet on the same cell and
71 the prey is not hidden. When the prey is on a hidden cell, the predator can enter that cell,
72 but does not interact with the prey. *Stochasticity* - Random numbers are used in
73 initialization of most variables, including coordinates of agents, the prey hiding property of
74 cells and the state variable 'gut-fill' of the predator. Stochasticity is also involved in the
75 moving direction of agents when random walk applies. Bernoulli-distributed random
76 numbers are drawn to determine the handling time for a prey item as time is a discrete
77 variable in this model. *Observation* In each simulation experiment, the number of prey
78 items eaten by the predator is recorded at each time step. At the end of each simulation
79 run, patch size, 'prey-hiding' degree (percentage of refuge cells), initial prey number, and
80 'prey-eaten' are recorded.

81 **1.6 Initialization**

82 Random values are used to initialize the spatial coordinates of all agents and choose prey
 83 hiding cells. All other initial parameters are listed in Table 1.

Table 1: Agents' state variables and parameters

sort	variables	unit	note
state variable	gut-fill	rate	randomly initialized
	still-handling	s	initially set to zero
	individual identify	number	
	prey-eaten	number	initially set to zero
calculated parameter	velocity	cm s^{-1}	
	rate of successful attack	rate	
	full gut	mg	allometrically calculated
	digestion rate	rate	
	handling time	s	

84 Most species traits regulating the processes described above are calculated following
 85 allometric rules. These include velocities of predator and prey, V [cm s^{-1}], Eq. (1a),
 86 (Peters, 1983), and the predator-specific traits: gut size, G [mg], Eq. (1b), (Ibarrola et al.,
 87 2012); digestion rate, D [mg s^{-1}], Eq. (1c), (Ibarrola et al., 2012); handling time, T_h [s],
 88 Eq. (1d), (modified from Rall et al., 2012); and rate of successful attack, S_a unitless, Eq.
 89 (1e), (Wahlström et al., 2000; data from Gergs and Ratte, 2009, and Gergs, 2011):

$$V = v_0 M^{a_v} \tag{1a}$$

$$G = g_0 M^{a_g} \tag{1b}$$

$$D = d_0 M^{a_d} \tag{1c}$$

$$T_h = h_0 M_p^{a_{h,p}} M_n^{a_{h,n}} \tag{1d}$$

$$S_a = a_0 m_0 \frac{M_p}{M_n} e^{1-m_0 \frac{M_p}{M_n}} \tag{1e}$$

90 where v_0 , g_0 and d_0 are constants, a_v , a_g , and a_d are the allometric exponents, and M is the
 91 body mass of the corresponding individual. Subscripts $_p$ and $_n$ of M indicate predator and
 92 prey, respectively. As only few relevant studies were found on digestion mechanisms, we
 93 developed our own equation for attack success, Eq. (1e), inspired by Wahlström et al.
 94 (2000), in which a_0 is maximum attack success and m_0 is optimal prey-predator body mass
 95 ratio. Predator and prey also possess some state variables to assist their decision making
 96 and activities, i.e. ‘position’ for all individuals; ‘gut fullness’ and ‘still handling’ for the
 97 predator; ‘prey identity’ for prey.

98 1.7 Submodels

99 1. Prey move. This process is the first one for each time step. Prey individuals do
 100 random walks consecutively according to their identity number. A random direction is
 101 generated and position changes in two dimensions are calculated according to prey velocity.
 102 Before updating the actual coordinates, wall boundary conditions are considered, checking
 103 if values of the coordinates would be beyond the boundaries. If so, the value is set close to

104 the coordinate value of that edge but with a distance to the edge of 10^{-6} .

105 2. Digest. The state variable ‘gut-fill’ of the predator is subtracted by ‘digestion rate’
106 in this process. It is executed each time step even when the gut is already empty. If the
107 value of gut fill drops below zero, it is set to zero.

108 3. Handle prey. This process is executed under the condition that the state variable
109 ‘still-handling’ has a positive value. The value of ‘still-handling’ is reduced by one in this
110 process.

111 4. Move. The predator does a random walk. If the predator is satisfied, meaning
112 ‘gut-fill’ exceeds 0.6, this process is skipped.

113 5. Encounter and attack? Here, predator coordinates are checked only when it is
114 about to hunt (‘gut-fill’ < 0.6). If the predator is currently in a prey hiding cell, the
115 hunting process is forfeited. If it is not forfeited, the following actions are executed. 1)
116 Check potential prey, checking if there exists one prey item that is in the same cell as the
117 predator. Checking order follows the prey’s identity numbers. As soon as one prey fulfills
118 the condition, the checking is finished. 2) If there is a potential prey item, a random
119 number (ranging from 0 to 1) is generated and compared to ‘rate of successful attack’ to
120 decide if this prey flees.

121 6. Attack success? If the attacked prey does not flee (attack success), values of
122 ‘gut-fill’ and ‘still-handling’ will be increased by the amounts calculated from the prey
123 mass and ‘prey-eaten’ will be increased by one. As time is discrete in our model, a
124 Bernoulli-distributed random number is drawn to make sure that the value of
125 ‘still-handling’ is an integer and on average still satisfying the calculated handling time. If
126 the prey flees (attack unsuccessful), this time step ends.

127 7. Replace caught prey. If attack succeeds, the killed prey item (i.e. its identity

128 number) would be randomly given a set of spatial coordinates, but body mass never
129 changes.

130 8. Output data. Data are recorded immediately after each of 3600 time steps. The
131 number of prey eaten and relevant input values are recorded, i.e. body masses of agents,
132 patch size, percentage of prey hiding cells and initial number of prey items.

133 2 allometric handling time

134 We used data from Rall et al. (2012) to parameterize the equation for handling time (T_h ,
135 Eq. (2d)). We only selected the data for predation (parasitism excluded) and for short
136 experimental duration (≤ 10 minutes) to make the analysis. We fitted the 67 data points
137 to a linear mixed-effects model ('lme' in the package 'nlme' in R, Pinheiro et al. (2016); R
138 Core Team (2016)). To correct for differences between studies, study identity was used as a
139 random factor, and all the variables (explanatory variables, body masses of predator and
140 prey species and dependent variable handling time) were ln-transformed. The statistics
141 showed that the handling time increased with increasing prey mass and decreased with
142 increasing prey mass and decreases with increasing predator mass (Fig. 1, Table 2).

Table 2: Statistics for handling time ^a

	Estimate	S.E.	p-value
intercept	3.624	0.839	0.0001
ln.pred ^b	-0.330	0.059	<0.0001
ln.prey ^c	0.173	0.051	0.0013

^ahandling time is ln-transformed

^bln-transformed predator body mass

^cln-transformed prey body mass

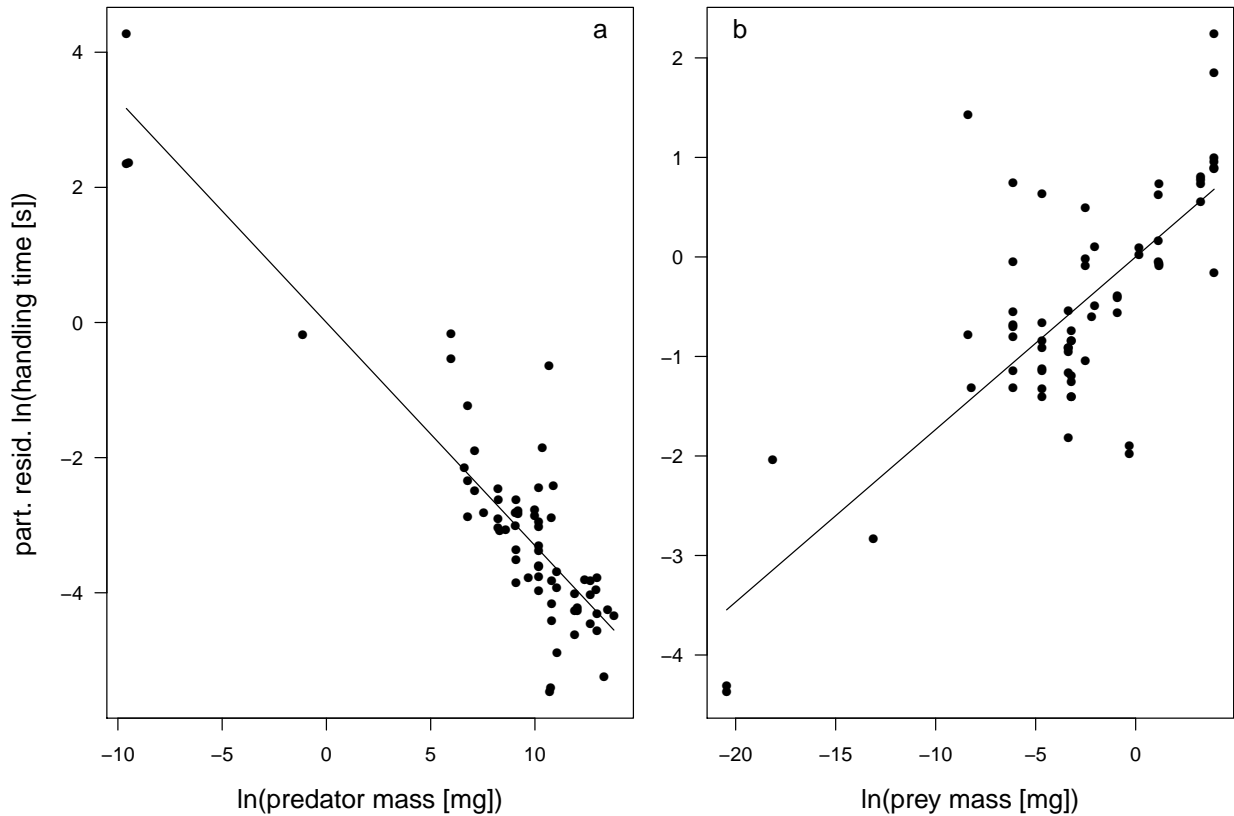


Figure 1: Fitting for handling time (T_h) shows that handling time decreases with increasing predator body mass and increases with increasing prey body mass. Data are ln-transformed before fitting. The partial residual of handling time is used as y-axis.

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