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Vlastimil Křivan. Evolutionary stability of optimal foraging: partial preferences in the diet and patch models. Journal of Theoretical Biology, 2010, 267 (4), pp.486. 10.1016/j.jtbi.2010.09.014 . hal-00637808

HAL Id: hal-00637808 https://hal.science/hal-00637808

Submitted on 3 Nov 2011

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Author's Accepted Manuscript

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PII:S0022-5193(10)00488-1DOI:doi:10.1016/j.jtbi.2010.09.014Reference:YJTBI6154

To appear in: Journal of Theoretical Biology

Received date:16 July 2010Revised date:8 September 2010Accepted date:9 September 2010



www.elsevier.com/locate/yjtbi

Cite this article as: Vlastimil Křivan, Evolutionary stability of optimal foraging: partial preferences in the diet and patch models, *Journal of Theoretical Biology*, doi:10.1016/j.jtbi.2010.09.014

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Evolutionary stability of optimal foraging: partial preferences in the diet and patch models

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Abstract

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In this article the patch and diet choice models of the optimal foraging theory are re-analyzed with respect to evolutionary stability of the optimal foraging strategies. In their original setting these fundamental models consider a single consumer only and the resulting fitness functions are both frequency and density independent. Such fitness function do not allow us to apply the classical game theoretical methods to study evolutionary stability of optimal foraging strategies for competing animals. In this article frequency and density dependent fitness functions of optimal foraging are derived by separation of time scales in an underlying population dynamical model and corresponding evolutionarily stable strategies are calculated. Contrary to the classical foraging models the results of the present article predict that partial preferences occur in optimal foraging strategies as a consequence of the ecological feedback of consumer preferences on consumer fitness. In the case of the patch occupation model these partial preferences correspond to the Ideal Free Distribution concept while in the case of the diet choice model they correspond to the partial inclusion of the less profitable prey type in predators diet.

Keywords: evolutionarily stable strategy, game theory, ideal free distribution, population growth, predator-prey models

7 1. Introduction

⁸ Optimal foraging theory (OFT; e. g., MacArthur and Pianka, 1966; Charnov, 1976; ⁹ Stephens and Krebs, 1986) assumes that organisms forage in such a way as to maximize ¹⁰ their fitness measured as energy intake rate. These models assume a homogeneous ¹¹ (fine-grained) environment with several resource types that a consumer encounters ¹² sequentially, and predict the optimal consumer diet. This line of research led to the ¹³ prey model (also called the "diet choice"; Charnov, 1976). This model assumes that ¹⁴ consumer fitness is measured as the average energy intake rate. The classical example

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of such a situation is the experiment with great tits where a single animal feeds on 15 two food types delivered on a conveyor belt (e. g., Krebs et al., 1977; Berec et al., 16 2003) which assures that prey are not depleted by predation. Thus, no interference 17 or competition for food between consumers is considered. A similar reasoning can be 18 applied to the case where a consumer forages in a heterogeneous environment consisting 19 of discrete foraging patches. In this case the consumer should move to the patch with 20 the highest payoff. These predictions are based on fitness functions that are both 21 density and frequency independent. They consider a single consumer and resource 22 densities that are not influenced by predation. Thus, they miss completely competition 23 between consumers that can be either due to interference, or due to exploitation of 24 resources. In fact, such models can be applied only to laboratory experiments with a 25 single consumer that does not influence resource density (e.g., a single great tit feeding 26 on two prev types delivered on the conveyor belt). 27

More realistic models must consider consumer competition (either interference com-28 petition, or exploitative competition). A density dependent approach in patchy envi-29 ronments led to the concept of the Ideal Free Distribution (IFD; Fretwell and Lucas, 30 1969). These authors assumed that patch payoff decreases with increasing number of 31 individuals in that patch (e.g., this captures interference among birds) and predicted 32 that under the IFD no individual can increase its fitness by changing its strategy 33 measured by residence times in different patches. Because under the IFD all patches 34 provide animals with the same fitness the question arises whether such a situation is 35 stable or not. Indeed, assuming a single mutant with an infinitely large resident pop-36 ulation that is distributed according to the IFD, fitness of that mutant is independent 37 and the same as is the fitness of resident animals. One approach to study stability in 38 such situations is to use the concept of evolutionarily stable strategies (Maynard Smith 39 and Price, 1973; Hofbauer and Sigmund, 1998). In fact it has been shown that the IFD 40 is an evolutionarily stable strategy of the habitat selection game (Křivan et al., 2008). 41 However, this concept of stability assumes that the fitness function is frequency depen-42 dent, i. e., it depends both on the mutant as well as on the resident strategy. Frequency 43 dependence allows us to predict whether a mutant strategy will increase in the resident 44 population or not. Here the idea is that the resident strategy sets the environment 45 (e. g., the resident consumer strategy sets the resource densities) and if a mutant strat-46 egy with a higher fitness exists then this strategy spreads, replaces residents, sets a 47 new environment and so on until an evolutionarily stable strategy (ESS) is reached. 48 Another approach that considers explicitly resource-consumer population dynam-49

ics together with optimal foraging was also used in theoretical ecology (reviewed in 50 Abrams, 2010). These models show how resident individuals that follow optimal for-51 aging strategy influence resource levels which, in turn, influence consumer numbers 52 and strategy. Although dynamical models capture the ecological feedback, they do 53 not provide us with information about the evolutionary stability of optimal foraging 54 strategies. To study this question it is necessary to study conditions under which 55 the optimal foraging strategy is resistant to mutant invasions. Studying evolution-56 ary stability in dynamic models of optimal foraging is more complicated than in static 57

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⁵⁸ models of behavioral ecology (such as the Hawk-Dove model, or the Prisoners dilemma;
⁵⁹ Maynard Smith and Price, 1973; Hofbauer and Sigmund, 1998) because it requires to
⁶⁰ analyze population dynamics for both resources, resident and mutant consumers (e. g.,
⁶¹ Křivan and Cressman, 2009; Cressman and Křivan, 2010).

To sum up, the problem is that some models of optimal foraging are described by fit-62 ness functions that are both frequency and density independent, and they do not allow 63 us to predict optimal foraging strategies when interference or resource depletion occur. 64 On the other hand, in the full dynamic setting that describes explicitly the effect of 65 resident strategies on resource densities, the resulting models are multidimensional and 66 difficult to analyze with respect to evolutionary stability of optimal strategies. My aim 67 is to show how frequency dependent fitness functions can be obtained from frequency 68 independent fitness functions, using a time scale argument. I start with a frequency 69 independent fitness function defined as the per capita population growth rate. Then I 70 consider resource–consumer population dynamics and I assume that resource dynamics 71 run on a faster time scale when compared with consumer population dynamics. This is 72 the case of many resource-consumer systems where resource generation time is shorter 73 than consumer generation time (e. g., many plants have short generation times rela-74 tive to mammalian herbivores). This time scale separation allows me to assume that 75 at each consumer density resources are at the corresponding population equilibrium. 76 Substituting this resource equilibrium to the consumer fitness function leads to a fre-77 quency dependent fitness function that can be analyzed from evolutionary perspective. 78 I apply this approach to diet and patch choice paradigms of optimal foraging theory. 79 In both cases this approach predicts emergence of partial preferences for alternative 80 prey/patch types. In the case of the patch model partial preferences describe the con-81 sumer Ideal Free Distribution (Fretwell and Lucas, 1969). Although partial preferences 82 are not predicted by the classical frequency independent models of optimal foraging, 83 they were observed in many foraging studies (for a review see Stephens and Krebs, 84 1986). Explanations for partial preferences range from incorrect classification of re-85 source and sampling by consumers (Krebs et al., 1977; Rechten et al., 1983), resource 86 crypsis (Erichsen et al., 1980), incorrect estimation of encounter rates with resource 87 by consumers (McNamara and Houston, 1987; Hirvonen et al., 1999), limited mem-88 ory capacity of consumers (Mangel and Roitberg, 1989: Bélisle and Cresswell, 1997). 89 and limited knowledge of the environment (Berec and Krivan, 2000). In this article 90 I will show another mechanism that explains emergence of partial preferences by the 91 ecological feedback of consumer preferences on their fitness. 92

⁹³ 2. The patch occupation model

I will consider an environment consisting of two foraging patches containing resources with abundance x_i in patch i (i = 1, 2). If e_i denotes the efficiency with which resources are converted to new consumers, λ_i is the consumer search rate for resources in patch i, and m_i is the consumer mortality rate in patch i then consumer fitness expressed as the instantaneous per capita population growth rate (cf. consumer per

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(3)

⁹⁹ capita population growth rate in model (3) below) is

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$$W = (e_1\lambda_1 x_1 - m_1)v_1 + (e_2\lambda_2 x_2 - m_2)v_2 \tag{1}$$

where v_i ($v_1 + v_2 = 1$) is the proportion of the time a consumer stays in patch *i*. Thus, v_i measures the consumer preference for patch *i*. Assuming that consumer preferences are adaptive, the optimal consumer strategy is to stay in patch 1 when patch payoff is higher there than in patch 2 and vice versa. Thus, the strategy that maximizes consumer fitness W is

$$v_1 = \begin{cases} 1 & \text{if } e_1 \lambda_1 x_1 - m_1 > e_2 \lambda_2 x_2 - m_2 \\ 0 & \text{if } e_1 \lambda_1 x_1 - m_1 < e_2 \lambda_2 x_2 - m_2. \end{cases}$$
(2)

This simple prediction states that consumers should move to the patch that provides 107 them with the highest fitness. It is clear that mutants using a different strategy obtain 108 a lower fitness and therefore, the above strategy is evolutionarily stable. However, 109 due to interference or resource depletion it is also obvious that as more and more 110 consumers move to a patch, this patch payoff must decrease, which is not captured by 111 maximization of fitness function W. To model effects of consumers on resources, I will 112 consider the following population dynamics (e. g., Fryxell and Lundberg, 1994, 1997; 113 Křivan, 1997; Křivan and Schmitz, 2003) 114

$$\frac{dx_1}{dt} = a_1 x_1 \left(1 - \frac{x_1}{L_1}\right) - v_1 \lambda_1 x_1 y$$

$$\frac{dx_2}{t} = a_2 x_2 \left(1 - \frac{x_2}{L_1}\right) - v_2 \lambda_2 x_2 y$$

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$$\frac{dx_2}{dt} = a_2 x_2 \left(1 - \frac{x_2}{L_2}\right) - v_2 \lambda_2 x_2 y
\frac{dy}{dt} = (e_1 \lambda_1 x_1 - m_1) v_1 y + (e_2 \lambda_2 x_2 - m_2) v_2 y$$

where y is consumer density, L_i is the resource i carrying capacity, and a_i is the resource per capita population growth rate. I stress here that consumer fitness (1) is indeed the per capita consumer population growth rate in model (3).

Using model (3) I derive a new frequency dependent fitness function. I consider a mutant with a strategy $\tilde{v} = (\tilde{v}_1, \tilde{v}_2)$ in a population of residents with a strategy $v = (v_1, v_2)$ and I re-write frequency independent fitness function (1) as a function of both mutant and resident strategies

$$W(\tilde{v}, v) = (e_1\lambda_1x_1 - m_1)\tilde{v}_1 + (e_2\lambda_2x_2 - m_2)\tilde{v}_2.$$

The dependence of this function on the resident strategy $(v = (v_1, v_2))$ is indirect through resource abundance that is set by the resident strategy in model (3). To make this dependence explicit I will assume that resource population dynamics run on a fast time scale when compared with consumer population dynamics so that for each consumer density resources relatively quickly reach their corresponding equilibrium densities. When residents use strategy $v = (v_1, v_2)$, the resource equilibrium at consumer density y, calculated from (3), is $x_i = L_i(1 - \lambda_i v_i y/a_i)$, (i = 1, 2). Substituting this density in W leads to a frequency dependent fitness function

$$W(\tilde{v}, v) = r_1 \tilde{v}_1 \left(1 - \frac{v_1 y}{K_1} \right) + r_2 \tilde{v}_2 \left(1 - \frac{v_2 y}{K_2} \right), \tag{4}$$

135 where

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$$r_i = e_i \lambda_i L_i - m_i$$
 and $K_i = \frac{a_i (e_i \lambda_i L_i - m_i)}{e_i L_i \lambda_i^2}$ (5)

are the per capita consumer population growth rate and the environmental carrying 137 capacity in patch i. It is an interesting observation that the new frequency dependent 138 fitness function is described by logistic consumer population growth in each patch. In 139 other words, if we assume that consumer population growth in each patch is logistic, 140 we would arrive immediately to fitness function (4). Evolutionarily stable strategy for 141 fitness function W was derived by Cressman et al. (2004) and I briefly recall main results 142 here. Assuming that $r_1 > r_2$, there are two possibilities. Either consumer density is 143 low $(y < (r_1 - r_2)K_1/r_1)$ in which case only the better patch 1 will be occupied, or 144 when consumer population is above the critical threshold then both patches will be 145 occupied by consumers, patch payoffs will be the same and the corresponding ESS is 146

$$v_1^* = \frac{K_1 r_2}{K_1 r_2 + K_2 r_1} + \frac{K_1 K_2 (r_1 - r_2)}{(K_1 r_2 + K_2 r_1) y}.$$
(6)

¹⁴⁸ Under this strategy no individual can increase its fitness by changing its strategy and ¹⁴⁹ this strategy cannot be invaded by individuals using a different strategy. This shows ¹⁵⁰ that when patch payoffs are the same the ESS depends on the consumer population ¹⁵¹ density. This ESS corresponds to the IFD of Fretwell and Lucas (1969). When con-¹⁵² sumers re-distribute instantaneously according to the IFD at each consumer density, ¹⁵³ consumer population growth is described by a piecewise logistic equation

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$$\frac{dy}{dt} = \begin{cases}
r_1 y \left(1 - \frac{y}{K_1}\right) & \text{if } y \leq \frac{(r_1 - r_2)K_1}{r_1} \\
\frac{r_1 r_2 (K_1 + K_2)}{K_2 r_1 + K_1 r_2} y \left(1 - \frac{y}{K_1 + K_2}\right) & \text{if } y > \frac{(r_1 - r_2)K_1}{r_1},
\end{cases} (7)$$

(Křivan and Sirot, 2002). The equilibrium of this model is $K_1 + K_2$ and at this equilibrium consumer preferences satisfy

 $\frac{v_1^*}{v_2^*} = \frac{K_1}{K_2}.$ (8)

These preferences correspond to the so called balanced dispersal under which no mutants can invade (McPeek and Holt, 1992; Padrón and Trevisan, 2006; DeAngelis et al.,
2007).

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Now I compare solutions of the resource-consumer model (3) (Figure 1, solid lines) 161 with optimal consumer strategy given by (2), with solutions of the single-species lo-162 gistic model (7) (Figure 1, dashed lines). Křivan and Schmitz (2003) showed that 163 for $L_i > m_i/(e_i\lambda_i)$, consumer population dynamics described by (3) converge on the 164 equilibrium density $K_1 + K_2$ at which the corresponding IFD is given by formula (8) 165 exactly as for model (7). This means that model (7) with optimal strategy (6) con-166 verges on the same equilibrium as the original resource-consumer model (3) (Figure 167 1). In addition, if resource population dynamics are much faster than consumer pop-168 ulation dynamics (Figure 1, left panel), population trajectories of the reduced model 169 (7) are close to trajectories of the original model (3). If both resource and consumer 170 population dynamics operate on comparable time scales (Figure 1, right panel) the 171 differences between trajectories get larger. 172

173 3. The diet choice model

Now I will consider the diet choice model. Assuming two prey types, consumer fitness in a fine-grained environment is

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$$\frac{e_1\lambda_1 u_1 x_1 + e_2\lambda_2 u_2 x_2}{1 + h_1\lambda_1 u_1 x_1 + h_2\lambda_2 u_2 x_2},\tag{9}$$

(Charnov, 1976). Here x_i is density of resource *i* in the environment, λ_i is the consumer 177 search rate for prey type i, u_i is the consumer preference for prey type i $(0 \le u_i \le 1)$, 178 h_i is the handling time, and e_i is the net energy gain obtained from prey type i. I stress 179 here that on the contrary to the patch model, the diet choice model does not assume 180 that the sum of strategies equals to one. Optimal foraging theory assumes that resource 181 densities are fixed, and maximization of (9), with respect to consumer preferences for 182 either resource, leads to the zero-one rule (Charnov, 1976). This rule predicts that 183 consumers will always forage on the most profitable resource type, i. e., the resource 184 type with the highest ratio e_i/h_i . In what follows I will assume that the resource 185 type 1 is more profitable than the alternative resource type 2 (i. e., $e_1/h_1 > e_2/h_2$). 186 Therefore, resource 1 will be always included in the consumer diet $(u_1 = 1)$ while the 187 second resource type will be included only if density of the first resource decreases 188 below the threshold density 189

$$x_1^* = \frac{e_2}{\lambda_1(e_1h_2 - e_2h_1)},\tag{10}$$

(e. g., Charnov, 1976; Stephens and Krebs, 1986). Consumer strategy is not uniquely predicted by the diet choice model when the more profitable prey density equals x_1^* , because under this condition any strategy will lead to the same fitness.

Now I will derive a frequency dependent fitness function and find the corresponding ESS. For this I consider ecological feedback between consumers and resources that is described by the following Rosenzweig–MacArthur predator-prey dynamics (Křivan

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and Schmitz, 2003)

$$\frac{dx_1}{dt} = a_1 x_1 \left(1 - \frac{x_1}{L_1} \right) - \frac{\lambda_1 x_1 y}{1 + h_1 \lambda_1 x_1 + h_2 \lambda_2 u_2 x_2}
\frac{dy}{dt} = \left(\frac{e_1 \lambda_1 x_1 + e_2 \lambda_2 u_2 x_2}{1 + h_1 \lambda_1 x_1 + h_2 \lambda_2 u_2 x_2} - m \right) y.$$
(11)

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This model considers population dynamics of the preferred resource (x_1) and consumers (y). It assumes that the alternative resource type density x_2 stays constant. This can be the case where there is an influx of alternative food resources to the system from outside sources (allochthonous resource; Huxel and McCann, 1998). Parameter u_2 describes consumer preferences for the alternative resource.

Fitness of a mutant consumer with strategy \tilde{u}_2 in a population of residents with strategy u_2 is

$$W(\tilde{u}_2, u_2) = \frac{e_1 \lambda_1 x_1 + e_2 \lambda_2 \tilde{u}_2 x_2}{1 + h_1 \lambda_1 x_1 + h_2 \lambda_2 \tilde{u}_2 x_2}$$
(12)

and I derive a frequency dependent fitness function as in the case of the patch model. For each fixed consumer strategy the resource 1 isocline (given by $dx_1/dt = 0$) is an upside down parabola (Figure 2). Assuming that resource dynamics run on a fast time scale when compared with consumer demography, trajectories move in the direction of the arrows in Figure 2 and they quickly reach the stable part of the resource isocline (solid line)

$$x_{1+} = \frac{-1 + h_1 L_1 \lambda_1 - h_2 u_2 x_2 \lambda_2 + H(u_2) \sqrt{1 - c(u_2)y}}{2h_1 \lambda_1}$$
(13)

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$$H(u_2) = 1 + h_1 L_1 \lambda_1 + h_2 x_2 u_2 \lambda_2, \quad c(u_2) = \frac{4h_1 L_1 \lambda_1^2}{a_1 H^2(u_2)}.$$

I remark that the resource isocline is defined only for consumer densities that satisfy $y < 1/c(u_2)$ which I will assume below. Substituting the equilibrium resource density x_{1+} in (12) leads to the frequency dependent fitness function

W(
$$\tilde{u}_2, u_2$$
) = $\frac{2h_1(e_1L_1\lambda_1 + e_2x_2\tilde{u}_2\lambda_2) - e_1H(u_2)(1 - \sqrt{1 - c(u_2)y})}{2h_1H(\tilde{u}_2) - h_1H(u_2)(1 - \sqrt{1 - c(u_2)y})}$. (14)

The selection gradient $\frac{\partial W}{\partial \tilde{u}_2}(\tilde{u}_2, u_2)$ predicts the direction of selective pressure (Vincent and Brown, 2005; Dercole and Rinaldi, 2008). When the gradient is positive, preference for the alternative prey type will increase, while if it is negative, preference will decrease. The gradient is zero at the singular strategy (Appendix A)

$$u_{2}^{*} = \frac{L_{1}\lambda_{1}^{2}(e_{1}h_{2} - e_{2}h_{1})}{a_{1}h_{2}\lambda_{2}x_{2}(L_{1}\lambda_{1}(e_{1}h_{2} - e_{2}h_{1}) - e_{2})}y - \frac{e_{1}}{\lambda_{2}x_{2}(e_{1}h_{2} - e_{2}h_{1})}$$
(15)

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provided this value is between zero and one, i. e., consumer density satisfies $y_1^* \le y \le y_2^*$ where

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$$y_1^* = \frac{a_1 e_1 h_2 (L_1 \lambda_1 (e_1 h_2 - e_2 h_1) - e_2)}{(e_2 h_1 - e_1 h_2)^2 L_1 \lambda_1^2}$$

228 and

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$$y_2^* = \frac{a_1 h_2 (L_1 \lambda_1 (e_1 h_2 - e_2 h_1) - e_2) (e_1 + (e_1 h_2 - e_2 h_1) x_2 \lambda_2)}{(e_2 h_1 - e_1 h_2)^2 L_1 \lambda_1^2}$$

from Appendix A. For consumer densities for which $y < y_1^*$, I set $u_2^* = 0$ and for $y > y_2^*$, 230 I set $u_2^* = 1$. Figure 3A shows dependence of the singular strategy on consumer density. 231 It is proved in Appendix A that the singular strategy is not only evolutionarily stable 232 but also convergence stable (Eshel, 1983; Eshel et al., 1997). This means that (1) 233 the singular strategy is resistant to invasion of a small number of mutants that use a 234 different strategy, and (2) it is also resistant to changes in the resident strategy (i. e., 235 when the resident strategy is slightly perturbed, it will return to the convergence stable 236 singular point). Strategies that are evolutionarily and convergence stable are thought 237 as the end point of evolution (Vincent and Brown, 2005; Dercole and Rinaldi, 2008). 238

Now I will analyze the singular consumer strategy as a function of consumer density.
If resource 1 carrying capacity is low so that

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$$L_1 < x_1^* \tag{16}$$

(where x_1^* is given by (10)) consumers will include the alternative prey type in their 242 diet $(u_2^* = 1)$ at all consumer densities, because the critical thresholds y_1^* and y_2^* are 243 negative. Because x_1^* is the critical threshold predicted by the optimal foraging theory, 244 below which the alternative prey type is included in the consumers' diet, the present 245 model predicts that at low environmental carrying capacities for the preferred prey type 246 consumers will behave as generalists. This is because at such low carrying capacities 247 the density of the more profitable prey type can never be higher than is the switching 248 threshold x_1^* . 249

When the carrying capacity of the preferred resource is higher than is the switching 250 density (i. e., the inequality in (16) is reversed), u_2^* is a piecewise linear function of the 251 consumer density y (Figure 3A). In this case at low consumer densities consumers will 252 behave as specialists feeding on the more profitable resource type only. As the number 253 of consumers increases resource 1 density will decrease and consumers will start to feed 254 on the alternative resource as well, and their preferences for this resource will increase 255 with consumer numbers. When at high numbers consumers will feed on the alternative 256 food items upon each encounter. This shows that when the consumer fitness reflects 257 frequency and density dependence given by the ecological feedback, partial preferences 258 do arise in the diet choice model. 259

Now I will consider consumer population dynamics. Consumer dynamics along the stable branch of the prey isocline are obtained by substituting expression for x_{1+} in the consumer equation in (11). This leads to the following consumer population growth 263 equation

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$$\frac{dy}{dt} = ry\left(1 - \frac{K(u_2)}{1 + \sqrt{1 - c(u_2)y}}\right)$$
(17)

where $r = \frac{e_1}{h_1} - m$ and $K(u_2) = 2\frac{e_1 + (e_1h_2 - e_2h_1)x_2\lambda_2u_2}{H(u_2)(e_1 - h_1m)}$ (Appendix B). Substituting for u_2 in model (17) the singular strategy (u_2^*) , describes consumer dynamics with optimal diet selection. For consumer densities for which the consumer singular strategy u_2^* is between 0 and 1, consumer population dynamics simplify to

$$\frac{dy}{dt} = y\frac{e_2 - h_2m}{h_2}$$

Depending on the sign of $e_2 - h_2 m$, consumer density either increases or decreases 270 when partial preferences occur. Because the above equation has no non-trivial pop-271 ulation equilibrium it follows that no population equilibrium where consumers would 272 show partial preferences for the alternative prey type exists. Consumer equilibria for 273 consumer densities that are higher than y_2^* (i. e., when $u_2^* = 1$) or smaller than y_1^* (i. e., when $u_2^* = 0$) are given by the interior equilibrium of (17) $y^* = \frac{K(u_2)(2-K(u_2))}{c(u_2)}$ 274 275 where I substitute 0 or 1 for u_2 (these are given explicitly in Appendix B). Figure 276 3B shows one such trajectory of model (17) driven by the consumer singular strategy 277 (solid lines). The corresponding resource 1 density is calculated from (13). Figure 3C 278 shows the corresponding consumer singular strategy. These results predict that at low 279 consumer densities where resource 1 density is near to its carrying capacity, consumers 280 specialize on the more profitable resource type. As consumers increase in numbers, the 281 preferred resource density decreases to the critical switching density (x_1^*) predicted by 282 the classical model of optimal foraging. From then on, consumer preferences for the 283 alternative prey type start to increase, keeping the preferred resource density at the 284 switching threshold for some time (i. e., times approx. between 5 and 14 in Figure 3B). 285 Because $e_2 - mh_2 > 0$ in Figure 3B, consumer density increases (it would decrease oth-286 erwise). Once the consumer preference reaches 1 (i. e., consumers attack the alternative 287 resource upon each encounter), consumers and resources tend to a stable equilibrium. 288 This general pattern of population dynamics is also clearly shown in Figure 3D where 289 the trajectory from panel B is projected on the consumer isocline (the solid dot de-290 notes the population equilibrium). Thus, in the region of consumer densities where 291 preferences for the less profitable resource type are between 0 and 1, consumers exhibit 292 partial preferences. This is a new result because the classical diet choice model does 293 not predict such a gradual shift in consumer preferences at the switching prey density 294 (Stephens and Krebs, 1986). Trajectories of model (11) with the optimal consumer 295 foraging strategy are shown in Figure 3B as dashed lines. It is clear that they converge 296 to the same equilibrium as trajectories of the single species model. Using a completely 297 different approach Křivan (1996) (see formula (32) there) calculated consumer strategy 298 at the switching threshold $x_1 = x_1^*$ (i. e., when the classical diet choice model does not 299 define consumer preference for the alternative prey type uniquely). It is an interesting 300

observation that the formula for partial preferences obtained there coincides with the singular strategy u_2^* .

303 4. Discussion

In this article I have shown how to derive a frequency dependent fitness function from a frequency independent fitness function by using a time-scale argument. The time scale argument assumes that resource population dynamics run on a faster time scale when compared with consumer population dynamics. Having a frequency dependent fitness function allowed me to study optimal foraging strategies for multiple consumers and depletable resources. In particular I showed that partial preferences for alternative resources arise in the diet and patch models of optimal foraging.

The diet and patch choice models are two paradigms of the optimal foraging theory 311 (MacArthur and Pianka, 1966; Emlen, 1966). They assume that per capita energy 312 intake rate is a proxy for consumer fitness that is maximized by adaptive consumer 313 foraging behavior. The diet choice model assumes that two or more resources are evenly 314 distributed in the environment and consumption of these resources does not influence 315 their numbers. This assumption is clearly violated in most natural systems. To comply 316 with these assumptions, e.g., conveyor belts that bring food directly to consumers were 317 used in experiments (Krebs et al., 1977; Berec et al., 2003). So what is missing in the 318 diet choice model is the link between consumer numbers and consumer strategy. This 319 link defines an ecological feedback mechanism: consumer foraging strategy influences 320 resource densities which, in turn, set the consumer strategy and density. To model this 321 ecological feedback mechanism some authors considered population dynamics together 322 with the optimal foraging models (e. g., Fryxell and Lundberg, 1994; Křivan, 1996; 323 Fryxell and Lundberg, 1997; Křivan, 1997; Křivan and Schmitz, 2003; Ma et al., 2003). 324 In these models it often happens that population dynamics tend to densities at which 325 models of optimal foraging do not predict the optimal strategy uniquely. For example, 326 in the patch model consumers will distribute between the two patches so that patch 327 payoffs will equalize (Krivan, 1997). Under this situation there is no selection against 328 mutants that can use any strategy, because individual fitness is independent of the 329 strategy. Similarly, dynamical models of diet choice drive periodically density of the 330 more profitable prev type to the threshold where optimal consumer diet is not uniquely 331 predicted (Křivan, 1996). In fact, this non-uniqueness causes fundamental problems 332 for the existence of solutions to these population models (Colombo and Křivan, 1993; 333 Křivan, 1996). 334

A simpler approach to study evolutionary stability is based on the concept of evolutionarily stable strategies (e. g., Maynard Smith and Price, 1973; Hofbauer and Sigmund, 1998; Cressman, 2003; Vincent and Brown, 2005). However, to apply methods of static game theory, the fitness function must be frequency dependent, i. e., it must allow us to measure the effect of mutants on the resident strategy. In this article I derived two frequency dependent fitness functions for patch and diet choice models. These fitness functions are based on the assumption that resource population dynamics

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run on a faster population time scale when compared with consumer population dy-342 namics. Using these fitness functions I derived the corresponding evolutionarily stable 343 strategies for the patch and the diet choice model. In both cases the corresponding 344 ESS predicts partial preferences for resources to occur. In the case of the patch model 345 separation of the resource and consumer time scales leads to consumer population 346 growth that is described by the logistic equation for which the evolutionarily stable 347 strategies were studied (e. g., Křivan and Sirot, 2002; Morris, 2003; Cressman and 348 Křivan, 2006). These works show that the evolutionarily stable strategy corresponds 349 with the IFD (Fretwell and Lucas, 1969). Similarly, when time scaling is applied to the 350 diet choice model the resulting consumer population growth equation seems to be new. 351 Moreover, partial preferences arise for a range of consumer densities in both models. 352 This is a new prediction because partial preferences were not predicted by the optimal 353 foraging theory (Charnov, 1976). In fact, my present analysis clearly shows that at 354 low consumer densities consumers should specialize on the more profitable prey/patch 355 type only. However, as consumer density increases, consumers also start to utilize the 356 alternative, less profitable prey/patch type with increasing strength. This prediction 357 should be easily tested using appropriate data on consumer preferences. 358

These predictions may have several consequences in population ecology. For example, if consumer preferences depend on consumer numbers, functional responses used in multiple species models should reflect this situation. Thus, they should depend not only on resource densities, but also on consumer densities (for a recent review of flexible foraging on the functional response see Abrams, 2010). Such a dependency could lead to a more mechanistic explanation of the effect of consumer density on the functional response.

366 Acknowledgments

The Institute of Entomology is funded by the Academy of Sciences of the Czech Republic (Z50070508). A part of this work was conducted while a Sabbatical Fellow at the National Institute for Mathematical and Biological Synthesis, an Institute sponsored by the National Science Foundation, the U.S. Department of Homeland Security, and the U.S. Department of Agriculture through NSF Award #EF-0832858, with additional support from The University of Tennessee, Knoxville.

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457 Appendix A: Evolutionary stability of the diet choice model

458 Fitness function (14) can be written as

$$W(\tilde{u}_2, u_2) = \frac{A\tilde{u}_2 + B}{C\tilde{u}_2 + D}$$

460 where

$$A = 2e_2h_1x_2\lambda_2, \quad B = 2h_1e_1L_1\lambda_1 - e_1H(u_2)(1 - \sqrt{1 - c(u_2)y}),$$

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$$C = 2h_1h_2x_2\lambda_2, \quad D = 2h_1(1+h_1L_1\lambda_1) - h_1H(u_2)(1-\sqrt{1-c(u_2)y}),$$

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$$c(u_2) = \frac{4h_1L_1\lambda_1^2}{a_1(1+h_1L_1\lambda_1+h_2x_2u_2\lambda_2)^2}, \quad H(u_2) = 1+h_1L_1\lambda_1+h_2x_2u_2\lambda_2.$$

First, I calculate the singular strategy and study its evolutionary stability. The gradient
of the fitness function with respect to the mutant strategy

$$\frac{\partial W}{\partial \tilde{u}_2}(\tilde{u}_2, u_2) = \frac{AD - BC}{(D + C\tilde{u}_2)^2}.$$

At the singular strategy this gradient when evaluated at $\tilde{u}_2 = u_2$ must be zero, i. e., ATO AD - BC = 0. When I substitute expressions for A, B, C, D and after some simplification I obtain the following equation

$$472 \quad \sqrt{1 - \frac{4h_1L_1y\lambda_1^2}{a_1(1+h_1L_1\lambda_1+h_2u_2x_2\lambda_2)^2}} = 1 - \frac{2h_1((e_1h_2-e_2h_1)L_1\lambda_1-e_2)}{(e_1h_2-e_2h_1)(1+h_1L_1\lambda_1+h_2u_2x_2\lambda_2)}.$$
 (18)

This equation posses a solution only provided the right hand side is between 0 and 1, i. e., when

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$$\frac{e_2}{\lambda_1(e_1h_2 - e_2h_1)} < L_1 < \frac{e_2h_1 + e_1h_2 + h_2(e_1h_2 - e_2h_1)u_2x_2\lambda_2}{h_1(e_1h_2 - e_2h_1)\lambda_1}.$$

Solving equation (18) yields the singular strategy (15). Of course, u_2^* must be between 477 0 and 1 which holds for consumer densities that satisfy $y_1^* \leq y \leq y_2^*$ where

$$y_1^* = \frac{a_1 e_1 h_2 (L_1 \lambda_1 (e_1 h_2 - e_2 h_1) - e_2)}{(e_2 h_1 - e_1 h_2)^2 L_1 \lambda_1^2}$$

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and

$$y_2^* = \frac{a_1 h_2 (L_1 \lambda_1 (e_1 h_2 - e_2 h_1) - e_2) (e_1 + (e_1 h_2 - e_2 h_1) x_2 \lambda_2)}{(e_2 h_1 - e_1 h_2)^2 L_1 \lambda_1^2}.$$

If consumer density is too low $(y < y_1^*)$, consumers will feed on the more profitable prey type only while at high densities $(y > y_2^*)$ they will feed on the alternative prey type upon each encounter. In particular, the carrying capacity for the more profitable prey type must be high enough, i. e.,

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$$L_1 > \frac{e_2}{\lambda_1(e_1h_2 - e_2h_1)}$$

for consumer partial preferences to arise. If the opposite inequality holds, fitness maximizes at $u_2 = 1$. I derived the singular strategy under the assumption that the right hand side of equation (18) is positive, because otherwise no singular solution exists. Substituting the singular strategy to the right hand side of (18) leads to expression

$$\frac{(e_2h_1 - e_1h_2)^2 L_1 y \lambda_1^2 - a_1h_1(e_2 + e_2h_1L_1\lambda_1 - e_1h_2L_1\lambda_1)^2}{(e_2h_1 - e_1h_2)^2 L_1 y \lambda_1^2 + a_1h_1(e_2 + e_2h_1L_1\lambda_1 - e_1h_2L_1\lambda_1)^2}$$

⁴⁹¹ that is positive provided

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$$y > \frac{a_1 h_1 (e_2 + e_2 h_1 L_1 \lambda_1 - e_1 h_2 L_1 \lambda_1)^2}{(e_2 h_1 - e_1 h_2)^2 L_1 \lambda_1^2}.$$
(19)

This condition must hold together with other constraints for the singular strategy to be between 0 and 1.

Because AD - BC = 0 at the singular strategy, $W(\tilde{u}_2, u_2^*)$ is independent of the mutant strategy \tilde{u}_2 and, after some calculations, $W(\tilde{u}_2, u_2^*) = e_2/h_2$ for all mutant strategies u_2 .

To study evolutionary stability of the singular strategy I need to check the local stability condition (Hofbauer and Sigmund, 1998) that asserts that mutants cannot spread in the resident population. The local stability condition requires that

$$W(u_2^*, u_2) > W(u_2, u_2) \tag{20}$$

for every $u_2 \neq u_2^*$ in a neighborhood of the singular strategy u_2^* . Let $g(u_2) = W(u_2^*, u_2) - W(u_2, u_2)$. Thus, $g(u_2^*) = 0$ and with a help of some computer algebra package (e. g., Mathematica) it is easy to show that provided inequality (19) holds, $g'(u_2^*) = 0$ and

$$g''(u_2^*) = \frac{2a_1^3h_2x_2^2(e_2 + (e_2h_1 - e_1h_2)L_1\lambda_1)^4\lambda_2^2}{(e_1h_2 - e_2h_1)L_1^2y^2\lambda_1^4((e_2h_1 - e_1h_2)^2L_1y\lambda_1^2 - a_1h_1(e_2 + (e_2h_1 - e_1h_2)L_1\lambda_1)^2)} > 0.$$

Thus, $g(u_2) > 0$ in a neighborhood of the singular strategy $(u_2 \neq u_2^*)$ and the local ESS condition (20) holds.

Second, I will prove that the singular strategy is also continuously stable. This follows from the fact (Eshel, 1983) that

$$\frac{\partial^2 W(u_2^*, u_2^*)}{\partial \tilde{u}_2^2} + \frac{\partial^2 W(u_2^*, u_2^*)}{\partial \tilde{u}_2 \partial u_2} =$$

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$$\frac{a_1^3 h_2 x_2^2 (e_2 + e_2 h_1 L_1 \lambda_1 - e_1 h_2 L_1 \lambda_1)^4 \lambda_2^2}{(e_1 h_2 - e_2 h_1) L_1^2 y^2 \lambda_1^4 (-(e_2 h_1 - e_1 h_2)^2 L_1 y \lambda_1^2 + a_1 h_1 (e_2 + e_2 h_1 L_1 \lambda_1 - e_1 h_2 L_1 \lambda_1)^2)} < 0$$

⁵¹³ for consumer densities that satisfy (19).

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⁵¹⁴ Appendix B: Derivation of model (17).

Substituting x_{1+} given by (13) in the right hand side of the consumer population growth equation (11) leads to

$$\frac{dy}{dt} = \frac{y(A + (e_1 - h_1m)\sqrt{-4h_1L_1y\lambda_1^2 + a_1H^2})}{h_1(\sqrt{a_1}H + \sqrt{-4h_1L_1y\lambda_1^2 + a_1H^2})}$$

518 where

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$$A = \sqrt{a_1}(-e_1 - h_1m + e_1h_1L_1\lambda_1 - h_1^2L_1m\lambda_1 + (2e_2h_1 - h_2(e_1 + h_1m))u_2x_2\lambda_2),$$

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$$H = 1 + h_1L_1\lambda_1 + h_2x_2u_2\lambda_2.$$

Extending the above fraction by expression $h_1(\sqrt{a_1}H - \sqrt{-4h_1L_1y\lambda_1^2 + a_1H^2})$ leads to

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$$\frac{dy}{dt} = \frac{(e_1 - h_1 m)y}{h_1} + \frac{\sqrt{a_1}(u_2 x_2 \lambda_2 (e_2 h_1 - e_1 h_2) - e_1)(\sqrt{a_1} H - \sqrt{a_1 H^2 - 4h_1 L_1 y \lambda_1^2})}{2h_1^2 L_1 \lambda_1^2}$$

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$$= \frac{(e_1 - h_1 m)y}{h_1} - \frac{2\sqrt{a_1}y(e_1 + u_2x_2\lambda_2(e_1h_2 - e_2h_1))}{h_1(\sqrt{a_1}H + \sqrt{a_1H^2 - 4h_1L_1y\lambda_1^2})}$$

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$$=\frac{(e_1-h_1m)y}{h_1}\left(1-\frac{2h_1\sqrt{a_1}(e_1+u_2x_2\lambda_2(e_1h_2-e_2h_1))}{(e_1-h_1m)h_1\sqrt{a_1}H(1+\sqrt{1-4h_1L_1y\lambda_1^2/(a_1H^2)})}\right)$$

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$$= ry\left(1 - \frac{K}{1 + \sqrt{1 - cy}}\right) \tag{21}$$

where $r = \frac{e_1}{h_1} - m$, $K = 2\frac{e_1 + u_2\lambda_2 x_2(e_1h_2 - e_2h_1)}{(e_1 - h_1m)H}$ and $c = \frac{4h_1L_1\lambda_1^2}{a_1H^2}$. There exists a non-zero equilibrium of the single species consumer model (21)

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$$y^* = \frac{K(2-K)}{c} = \frac{a_1(e_1 + (e_1h_2 - e_2h_1)u_2x_2\lambda_2)(L_1(e_1 - h_1m)\lambda_1 - m + (e_2 - h_2m)u_2x_2\lambda_2)}{L_1\lambda_1^2(e_1 - h_1m)^2}$$

which is exactly the same as the interior consumer equilibrium of model (11). For y^* to be an equilibrium, it must be positive, i. e., 0 < K < 2. This condition is equivalent to

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$$h_1 < e_1/m, \quad L_1 > \frac{m - (e_2 - h_2 m) u_2 x_2 \lambda_2}{\lambda_1 (e_1 - m h_1)}.$$

Equilibrium y^* is asymptotically stable when the sign of the derivative of the right hand side of (21) evaluated at this equilibrium

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$$J = \frac{d}{dy} \left(ry \left(1 - \frac{K}{1 + \sqrt{1 - cy}} \right) \right) |_{y=y^*} = \frac{r(K-2)}{2(K-1)}$$

is negative, i. e., 1 < K < 2. I remark that condition 1 < K is equivalent to

which is the condition for the interior equilibrium of the resource-consumer model (11), to be stable. For larger values of the environmental carrying capacities the interior equilibrium is unstable and a stable limit cycle arises in the Rosenzweig-MacArthur model (11) (Hofbauer and Sigmund, 1998).

 $L_1 < \frac{e_1 + h_1 m + (-2e_2h_1 + h_2(e_1 + h_1m))u_2 x_2 \lambda_2}{h_1(e_1 - h_1m)\lambda_1}$

The above analysis assumed fixed consumer preference for the alternative resource. Now I will study population dynamics (17) driven by the singular strategy. Substituting singular strategy (15) in population dynamics (17) I get

$$\frac{dy}{dt} = y \frac{(e_2 - h_2 m)}{h_2}$$

for consumer densities satisfying $y_1^* < y(t) < y_2^*$ and (19). For $h_2 < e_2/m$ consumer densities increase, while for larger handling times they decrease. No nontrivial equilibrium exists. If $y \ge y_2^*$ then $u_2 = 1$ and population dynamics on the stable manifold are obtained by substituting $u_2 = 1$ to (17). These population dynamics have equilibrium

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$$y_{eq}^{1} = \frac{a_{1}(e_{1} + (e_{1}h_{2} - e_{2}h_{1})x_{2}\lambda_{2})(e_{1}L_{1}\lambda_{1} + e_{2}x_{2}\lambda_{2} - m(1 + h_{1}L_{1}\lambda_{1} + h_{2}x_{2}\lambda_{2}))}{L_{1}(e_{1} - h_{1}m)^{2}\lambda_{1}^{2}}$$

Similarly when $y < y_1^*$, $u_2 = 0$ and population dynamics on the stable manifold are obtained by substituting $u_2 = 0$ to (17). These population dynamics have equilibrium

$$y_{eq}^{0} = \frac{a_{1}e_{1}(e_{1}L_{1}\lambda_{1} - m(1 + h_{1}L_{1}\lambda_{1}))}{L_{1}(e_{1} - h_{1}m)^{2}\lambda_{1}^{2}}$$

⁵⁵⁸ Figures caption.

Figure 1: Solutions of the patch model (3) (solid line) where consumers follow 559 the optimal foraging strategy given by (2), and solutions of the single-species model 560 (7)(dashed line). Resource densities for model (7) are given by $x_i = L_i(1 - y(t)\lambda_i v_i/a_i)$ 561 with the optimal strategy v_i given by (6). The left panel (A,B,C) assumes that con-562 sumer demographic parameters ($e_1 = 0.015$, $e_2 = 0.01$, $m_1 = m_2 = 0.02$) are much 563 smaller when compared to resource parameters. This discrepancy causes resource pop-564 ulation dynamics to run on a fast time scale when compared with consumer population 565 dynamics. The right panel (D,E,F) assumes more similar time scales for both resource 566 and consumer dynamics $(e_1 = 0.15, e_2 = 0.1, m_1 = m_2 = 0.2)$. Other parameters used 567 in simulations: $L_1 = L_2 = 10$, $a_1 = 1.5$, $a_2 = 0.5$, $\lambda_1 = \lambda_2 = 1$. 568

Figure 2: Resource isocline of model (11). Only the solid part of the isocline is stable provided resource 1 population dynamics are fast when compared to consumer population dynamics. Arrows indicate direction of trajectories.

Figure 3: Panel A shows consumer preference (15) for the alternative prey type in 572 the diet choice model as a function of consumer density. Panel B compares a trajectory 573 of the resource–consumer model (11) (dashed lines) where consumers follow predictions 574 of the classical prey model (i. e., zero-one rule) with a trajectory of the reduced model 575 (17) (solid lines). The corresponding trajectory for resources is given by (13). Panel C 576 shows the corresponding consumer optimal strategy. Panel D shows the solid trajectory 577 from panel B in the resource 1-consumer preference-consumer density phase space. 578 Parameters: $a_1 = 1.2, h_1 = h_2 = 0.2, e_1 = 0.2, e_2 = 0.05, \lambda_1 = \lambda_2 = 0.5, x_2 = 8,$ 579 $m = 0.2, L_1 = 10.$ 580 Accepter



Figure 1:

u_{0}^{12}

ACCEPTED MANUSCRIPT

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Figure 3: