

## COMPETITION BETWEEN A NONALLELOPATHIC PHYTOPLANKTON AND AN ALLELOPATHIC PHYTOPLANKTON SPECIES UNDER PREDATION

JEAN-JACQUES KENGWOUNG-KEUMO

Department of Mathematical Sciences, Cameron University  
2800 West Gore Boulevard, Lawton, OK 73505, USA

(Communicated by Hermann J. Eberl)

**ABSTRACT.** We propose a model of two-species competition in the chemostat for a single growth-limiting, nonreproducing resource that extends that of Roy [38]. The response functions are specified to be Michaelis-Menten, and there is no predation in Roy's work. Our model generalizes Roy's model to general uptake functions. The competition is exploitative so that species compete by decreasing the common pool of resources. The model also allows allelopathic effects of one toxin-producing species, both on itself (autotoxicity) and on its nontoxic competitor (phytotoxicity). We show that a stable coexistence equilibrium exists as long as (a) there are allelopathic effects and (b) the input nutrient concentration is above a critical value. The model is reconsidered under instantaneous nutrient recycling. We further extend this work to include a zooplankton species as a fourth interacting component to study the impact of predation on the ecosystem. The zooplankton species is allowed to feed only on the two phytoplankton species which are its perfectly substitutable resources. Each of the models is analyzed for boundedness, equilibria, stability, and uniform persistence (or permanence). Each model structure fits very well with some harmful algal bloom observations where the phytoplankton assemblage can be envisioned in two compartments, toxin producing and non-toxic. The *Prymnesium parvum* literature, where the suppressing effects of allelochemicals are quite pronounced, is a classic example. This work advances knowledge in an area of research becoming ever more important, which is understanding the functioning of allelopathy in food webs.

**1. Introduction.** The term “plankton” is used to describe freely-floating and weakly-swimming marine and freshwater organisms. It was coined by the German scientist Victor Hensen in 1887 (Thurman [46]). Plankton are divided into broad functional groups, among them phytoplankton that live near the surface of the water where there is sufficient light to support photosynthesis, and zooplankton that feed on other plankton. The microscopic and unicellular plants, phytoplankton, are consumed by zooplankton, the animals, which in turn are eaten by larger organisms. Plankton are at the base of the food chain in the aquatic environment, and are responsible for much of the oxygen present in the Earth's atmosphere: half of the total amount produced by all plant life. They also absorb carbon dioxide from their surrounding environment. The highly diverse nature of phytoplankton

---

2010 *Mathematics Subject Classification.* 91B74, 97M10, 62P12.

*Key words and phrases.* Allelopathy, phytoplankton, zooplankton, chemostat, predation, nutrient recycling.

communities seems to contradict the competitive exclusion principle, which states that when two species compete for the same resource, only one will survive. The modeling of plankton populations is an essential tool in improving our understanding of the physical and biological processes that contribute to the complexity of these systems.

The term *allelopathy* was first coined by the Austrian plant physiologist Molisch [29] to explain the effect of ethylene on fruit ripening. Rice [36] defines allelopathy as the effects of one plant (including micro-organisms) on the growth of another plant through the release of chemical compounds (called *allelochemicals* by Whittaker and Feeny [47]) into the environment. At the 1996 meeting of the International Allelopathy Society (IAS), *allelopathy* was redefined as any process involving secondary metabolites produced by plants, algae, bacteria, and fungi that influences the growth and development of agricultural and biological systems.

Mathematical modeling of plankton populations goes back to Riley [37] and Hallam [14, 15, 16]. The first mathematical model to represent allelopathic interactions between competing species was introduced by Maynard-Smith [27]. Improvements and refinements of Maynard-Smith's model have followed several directions (see for examples Nakamaru and Iwasa [32], An et al. [1] and references therein).

The two functions of allelopathy are autotoxicity (where the toxin affects the growth of the toxic species itself) and phytotoxicity (the effect of toxin on the growth of another species). We assume that the phytotoxic term depends upon the product of square of concentration of non-toxic species with square of concentration of toxic species in accordance with Solé *et al.* [44]. In addition, we suppose that the autotoxic term is a linear function of the concentration of the toxic species as per Sinkkonen [42].

To better understand the dynamics of plankton, one should not neglect the interactions of zooplankton. Many authors have worked in this direction. Edwards and Brindley [11] investigated the bifurcational structure of a simple plankton model with zooplankton mortality modeled by  $-cZ^m$ ,  $1 \leq m \leq 2$ . They showed explicitly how cycles can persist for  $1 < m < 2$ . In addition,  $m = 2$  does not preclude the existence of cycles or chaos. Edwards [12] examined the behavior of two nutrient-phytoplankton-zooplankton-detritus models to help understand the factors that most influence the dynamics of such models. He further showed that the addition of a detritus compartment has little impact on the nature of the qualitative dynamics that were found for the corresponding nutrient-phytoplankton-zooplankton model. Mukhopadhyay and Bhattacharyya [31] examined a model of nutrient-phytoplankton-zooplankton interaction with spatial heterogeneity. They proved that phytoplankton species with low diffusivity and zooplankton functional response with half-saturation constant can control algal blooms. Ruan [41] studied plankton nutrient models with both instantaneous and delayed nutrient recycling. He successively chose the nutrient input concentration and the maximal zooplankton ingestion rate as bifurcation parameters to show that the positive equilibrium loses its stability via a Hopf bifurcation as these parameters are varied through respective critical values. Jang and Baglama [22] explored nutrient-phytoplankton interaction with both instantaneous and delayed nutrient recycling and zooplankton mortality modeled by  $-cZ^2$ . Unlike other ecological models for which delays can destabilize the system (see Roy *et al.* [39] and Piotrowska *et al.* [34]), their numerical simulations suggested that delayed nutrient recycling can actually stabilize the nutrient-phytoplankton system. Chakraborty and Chattopadhyay [9] proposed

and analyzed four models of nutrient-phytoplankton-zooplankton populations to observe the dynamics of such models in the presence of additional food. Here the phytoplankton are toxic to the zooplankton population. However, little has been done in modeling competition, allelopathy, predation, and instantaneous nutrient recycling in the same food chain. The model we propose incorporates competition, allelopathy, predation, and instantaneous nutrient recycling.

With the use of mathematical models, Roy [38] demonstrates theoretically that the stable coexistence of two species competing for a single nutrient in a homogeneous medium would be possible provided one of the two species has a sufficiently strong allelopathic effect on the other. The uptake functions are specified to be Michaelis-Menten in Roy's model. The contents of this paper are largely devoted to extending the results of Roy [38]. More specifically, we extend the nutrient-nontoxic phytoplankton-toxic phytoplankton model of Roy [38] to general uptake functions and include a zooplankton species as a fourth interacting component. We carefully structure this work in a way that presents the incremental effects of adding the increased complexity and realism to the model.

This paper is organized as follows. In the next section, we describe the main model. We then present some preliminary results, and study existence and local stability of steady states. Ecological interpretations of inequalities are followed by some global results. We further extend this work to include a zooplankton species to study the impact of predation on the competition-allelopathy model. The zooplankton species predate only on the two phytoplankton species which are its perfectly substitutable resources. We reconsider each of the models under the effects of instantaneous nutrient recycling. We utilized Matlab and Mathematica Version 10 to run extended simulations to support our analytical findings.

**2. The Competition-Allelopathy model.** We will consider population interactions in a chemostat environment. For a detailed description of the chemostat and its application in biology and ecology, the reader is referred to Hsu *et al.* [20], Smith and Waltman [43].

Two-species compete exploitatively for a single nonreproducing resource. Our model also incorporates allelopathic effects, and can be written

$$\begin{aligned}\frac{dN}{dt} &= (N_0 - N)D - \frac{1}{\gamma_1}f_1(N)P_1 - \frac{1}{\gamma_2}f_2(N)P_2, \\ \frac{dP_1}{dt} &= (f_1(N) - m_1)P_1 - \gamma P_1^2 P_2^2, \\ \frac{dP_2}{dt} &= (f_2(N) - m_2)P_2, \\ N(0) &> 0, P_1(0) > 0, P_2(0) > 0.\end{aligned}\tag{1}$$

In these equations  $P_i(t)$  is the biomass of the  $i^{\text{th}}$  population of phytoplankton in the culture vessel at time  $t$ ,  $i = 1, 2$ . Population  $P_1$  is assumed to be nontoxic, while population  $P_2$  is assumed to be toxic. The concentration of the nonreproducing resource in the culture vessel at time  $t$  is denoted by  $N(t)$ , while  $N_0$  is the concentration of resource  $N$  in the feed vessel.

The removal rate  $m_1$  of nontoxic phytoplankton  $P_1$  is the sum of washout rate  $D$  and the specific death rate  $\epsilon_1$ , so that,  $m_1 = D + \epsilon_1$ . The removal rate  $m_2$  of toxic phytoplankton  $P_2$  is the sum of washout rate  $D$ , the specific death rate  $\epsilon_2$ , and the autotoxic coefficient  $a_2$ , so that,  $m_2 = D + \epsilon_2 + a_2$  as per Sinkkonen [42].

As in Wolkowicz and Lu [48], it is interesting to note that the analysis of the model requires no assumptions on the signs of the  $\epsilon_i$ 's and  $a_2$ , as long as the  $m_i$ 's all remain positive. This leaves the  $\epsilon_i$ 's and  $a_2$  open to other interpretations. For instance, a negative  $\epsilon_i$  describes an additional food source for the  $i^{\text{th}}$  population of phytoplankton while a positive  $\epsilon_i$  accounts for further deleterious effects (such as sinking and mixing) on the  $i^{\text{th}}$  population of phytoplankton. Finally, a zero  $\epsilon_i$  means that there is no intrinsic death of the  $i^{\text{th}}$  population of phytoplankton. A negative, zero, and positive  $a_2$  indicate respectively stimulatory effects, no effects, and inhibitory effects of toxins produced by  $P_2$  on the growth of conspecifics.

As in Solé *et al.* [44], we express the phytotoxic interactions as  $\gamma P_1^2 P_2^2$ , where  $\gamma$  denotes the phytotoxic coefficient. In system (1) the response functions  $f_i(N)$  represent the per capita rate of conversion of nutrient to biomass of population  $P_i$  per unit of population  $P_i$  as a function of the concentration of nutrient  $N$ . We assume that the rate of conversion of nutrient to biomass is proportional to the amount of nutrient consumed, so that the consumption rate of resource  $N$  per unit of population  $P_i$  is of the form  $\frac{1}{\gamma_i} f_i(N)$ , where  $\gamma_i$  is the growth yield constant (number of phytoplankton per unit of nutrient). We make the following assumptions concerning the response functions  $f_i$ :

$$f_i : \mathbf{R}_+ \longrightarrow \mathbf{R}_+, f_i \text{ is continuously differentiable,} \quad (2)$$

$$f_i(0) = 0, f_i'(N) > 0 \text{ for all } N \geq 0. \quad (3)$$

The break-even concentration for population  $P_i$  on nutrient  $N$  is obtained by setting  $\frac{dP_i}{dt} = 0 = f_i(N) - m_i$  and solving for  $N$ . By the monotonicity assumptions, the solution  $\lambda_i$  is a uniquely defined positive extended real number provided we assume  $\lambda_i = \infty$  if  $f_i(N) < m_i$  for all  $N \geq 0$ . Finally, let  $\mu_i$  denote the maximal growth rate of population  $P_i$  on resource  $N$ , so that

$$\lim_{N \rightarrow \infty} f_i(N) = \mu_i.$$

Lotka-Volterra kinetics (or Holling type I), Michaelis-Menten kinetics (or Holling type II), and sigmoidal kinetics (Holling type III or multiple saturation dynamics) are prototypes of response functions  $f_i$  found in the literature (Aris and Humphrey [3], Boon and Laudelout [6], Edwards [12], Jost et al. [23], Monod [30], Ruan [40], Wolkowicz and Lu [48], Yang and Humphrey [49]).

The half-saturation constant  $K_i$  of the  $i^{\text{th}}$  phytoplankton species for nutrient is given by  $f_i(K_i) = \mu_i/2$  and so represents the resource concentration supporting growth at half the maximal growth rate. Half-saturation constants and maximal growth rates can be measured experimentally (Hansen and Hubbell [17]). System (1) was considered by Roy [38] under the assumption that the response functions  $f_i$  are Michaelis-Menten and the yield constants  $\gamma_i$  equal 1.

**2.1. Preliminary results.** The first lemma is a statement that solutions of (1) are positive and bounded. These are minimal requirements for a reasonable model of the chemostat.

**Lemma 2.1.** (a) *Solutions of model (1) are positive and bounded.*

(b) *Given any  $\delta > 0$ , for all solutions  $N(t)$  of (1)  $N(t) \leq N_0 + \delta$ , for all sufficiently large  $t$ .*

(c) *If there exists a  $t_0 \geq 0$  such that  $N(t_0) \leq N_0$ , then  $N(t) < N_0$  for all  $t > t_0$ .*

*Proof of (a).* Assume there is a first time  $t_0 > 0$  such that  $N(t_0) = 0$  and  $N(t) > 0$  for all  $0 < t < t_0$ . Then  $\frac{dN}{dt}(t_0) \leq 0$ . However, by the first equation of model (1),  $\frac{dN}{dt}(t_0) = N_0D > 0$ , a contradiction.

We now show that  $P_i(t) > 0$  for all  $t, i = 1, 2$ . For  $i = 1$ , let  $t_1 = \min\{t > 0, P_1(t) = 0\}$  and

$$A_1 = \min \{f_1(N(t)) - m_1 - \gamma P_1(t)P_2^2(t), 0 \leq t \leq t_1\}.$$

Then for  $t \in [0, t_1]$ ,  $\frac{dP_1(t)}{dt} \geq A_1P_1(t)$ , so that,  $P_1(t) \geq P_1(0) \exp(A_1t) > 0$ . In particular,  $P_1(t_1) \geq P_1(0) \exp(A_1t_1)$ , a contradiction.

To show that  $P_2(t) > 0$  for all  $t$ , replace  $t_1$  with  $t_2 = \min\{t > 0 : P_2(t) = 0\}$  and  $A_1$  with  $A_2 = \min\{f_2(N(t)) - m_2, 0 \leq t \leq t_2\}$ , in the proof of  $P_1(t) > 0$  for all  $t$ .

To prove boundedness of solutions of model (1), define  $T(t) = N(t) + \frac{1}{\gamma_1}P_1(t) + \frac{1}{\gamma_2}P_2(t)$ . From (1) we have  $\frac{dT}{dt} \leq N_0D - D_0T$ , where  $D_0 = \min(D, m_1, m_2)$ . By Gronwall's lemma,  $T(t) \leq \frac{dN_0}{D_0} + (T(0) - \frac{dN_0}{D_0}) \exp(-D_0t)$ , and so  $N(t) + \frac{1}{\gamma_1}P_1(t) + \frac{1}{\gamma_2}P_2(t) \leq \frac{N_0d}{D_0}$  if  $T(0) < \frac{N_0D}{D_0}$  and  $N(t) + \frac{1}{\gamma_1}P_1(t) + \frac{1}{\gamma_2}P_2(t) \leq T(0)$  otherwise. Hence, by the positivity of solutions of model (1), all solutions of (1) are bounded.

*Proof of (b).* Let  $\delta > 0$  be given. From the first equation of (1) we have

$$\frac{dN}{dt} = (N_0 - N)D - \frac{1}{\gamma_1}f_1(N)P_1 - \frac{1}{\gamma_2}f_2(N)P_2 \leq (N_0 - N)D. \tag{4}$$

Hence,  $N(t) \leq N_0 + [N(0) - N_0] \exp(-tD)$  for all  $t$ . Since  $[N(0) - N_0] \exp(-tD)$  approaches 0 as  $t$  tends to infinity,  $N(t) \leq N_0 + \delta$  for all sufficiently large  $t$ .

*Proof of (c).* Suppose there exists a first time  $\hat{t} > t_0$  such that  $N(\hat{t}) = N_0$  and  $N(t) < N_0$  for all  $t_0 \leq t < \hat{t}$ . Then  $\frac{dN}{dt}(\hat{t}) \geq 0$ . However, by (1),  $\frac{dN}{dt}(\hat{t}) \leq -\sum_{i=1}^2 \frac{P_i(\hat{t})}{\gamma_i} f_i(N(\hat{t})) < 0$ , a contradiction.

The Fundamental Existence-Uniqueness Theorem (see, for example, Perko [33]) and Lemma 2.1(a) ensure that solutions of (1) exist uniquely for all time.

In the absence of allelopathic effects ( $\gamma = 0$ ), system (1) reduces to the model studied by Hsu [21] in the case of two competing species. As such, the system exhibits the competitive exclusion principle, which Hardin [18] states as “complete competitors cannot coexist”. That is, two species cannot coexist if they compete for a single resource available in growth-limiting amounts.

Of note, if  $\gamma = 0$  then only one species survives, the one with the lower breakeven concentration.

**2.2. Steady states: Existence and local stability.** Steady states of model (1) are solutions of:

$$\frac{dN}{dt} = \frac{dP_1}{dt} = \frac{dP_2}{dt} = 0.$$

In what follows,  $\lambda_i$  is the unique positive solution of the equation  $f_i(N) = m_i$ . Three of the steady states are readily identified and are given by:

$$E_0 = (N_0, 0, 0), E_{\lambda_1} = (\lambda_1, \bar{P}_1, 0), \text{ and } E_{\lambda_2} = (\lambda_2, 0, \bar{P}_2), \text{ where } \bar{P}_i = \gamma_i(N_0 - \lambda_i)D/m_i, \text{ for } i = 1, 2.$$

We say that a steady state does not exist if any one of its components is negative.  $E_0$  always exists, whereas for each  $i \in \{1, 2\}$  a necessary and sufficient condition on

the parameters for feasibility of  $E_{\lambda_i}$  is  $N_0 > \lambda_i$ . Note that when  $N_0 = \lambda_i$ ,  $E_0$  and  $E_{\lambda_i}$  coalesce.

If any other equilibria exist, they must be interior equilibria (for which  $N$ ,  $P_1$ , and  $P_2$  are all positive). By the third equation of (1), we must have  $\hat{N} = \lambda_2$ , so that an interior equilibrium  $\hat{E} = (\lambda_2, \hat{P}_1, \hat{P}_2)$  of (1) corresponds to solutions  $(\hat{P}_1, \hat{P}_2) \in \text{Int}(\mathbf{R}_+^2)$  of the system

$$\begin{aligned} \frac{1}{\gamma_1} f_1(\lambda_2) P_1 + \frac{1}{\gamma_2} m_2 P_2 &= (N_0 - \lambda_2) D, \\ \gamma P_1 P_2^2 &= f_1(\lambda_2) - m_1. \end{aligned} \quad (5)$$

Note that there is no interior equilibrium when  $\lambda_1 \geq \lambda_2$ . Using a method similar to that of Roy [38] we establish the existence of zero, one or two interior equilibria as a function of the value of the input nutrient concentration  $N_0$ . The second equation of system (5) defines a curve  $F_2(P_2) = (f_1(\lambda_2) - m_1)/\gamma P_2^2$  in the first quadrant convex to the origin and satisfying  $\lim_{P_2 \rightarrow \infty} F_2(P_2) = 0$  and  $\lim_{P_2 \rightarrow 0^+} F_2(P_2) = \infty$ . The first equation of system (5) gives a straight line in the first quadrant with slope  $-\gamma_2 f_1(\lambda_2)/\gamma_1 m_2$  and  $P_2$ -intercept  $\gamma_2(N_0 - \lambda_2)D/m_2$ . These curves may or may not intersect in the first quadrant. If  $N_0$  is too small, the two curves do not intersect and model (1) does not have an interior equilibrium point. As  $N_0$  is increased, there is a critical value  $N_0^c$  for which the straight line is tangent to the curve, and model (1) has precisely one equilibrium given by  $\tilde{E} = (\lambda_2, \tilde{P}_1, \tilde{P}_2)$ , where  $\tilde{P}_1 = \gamma_1(N_0 - \lambda_2)D/3f_1(\lambda_2)$ , and  $\tilde{P}_2 = 2\gamma_2(N_0 - \lambda_2)D/3m_2$ . The critical value is computed using system (5), and is given by

$$N_0^c = \lambda_2 + \sqrt[3]{\frac{27 f_1(\lambda_2) m_2^2 (f_1(\lambda_2) - m_1)}{4 \gamma_1 \gamma_2^2 D^3 \gamma}}. \quad (6)$$

As  $N_0$  is increased beyond  $N_0^c$ , the straight line intersects the curve in two points and (1) has two interior equilibria  $\hat{E}' = (\lambda_2, \hat{P}'_1, \hat{P}'_2)$  and  $\hat{E}'' = (\lambda_2, \hat{P}''_1, \hat{P}''_2)$ , where  $0 < \hat{P}'_1 < \tilde{P}_1$ ,  $\tilde{P}_2 < \hat{P}'_2 < 3\tilde{P}_2/2$ ,  $\tilde{P}_1 < \hat{P}''_1 < 3\tilde{P}_1$ , and  $0 < \hat{P}''_2 < \tilde{P}_2$  (see Figure 1). Recall  $\tilde{P}_1 = \gamma_1(N_0 - \lambda_2)D/3f_1(\lambda_2)$ , and  $\tilde{P}_2 = 2\gamma_2(N_0 - \lambda_2)D/3m_2$  are respectively the  $P_1$ - and  $P_2$ -coordinates of the unique interior equilibrium  $\tilde{E}$  when  $N_0 = N_0^c$ .

We now investigate the local stability properties of (1) through an examination of the linearized system about each of the equilibria. The variational matrix of (1), denoted by  $V(N, P_1, P_2)$ , is given by

$$\begin{pmatrix} -D - \frac{P_1}{\gamma_1} f'_1(N) - \frac{P_2}{\gamma_2} f'_2(N) & -\frac{f_1(N)}{\gamma_1} & -\frac{f_2(N)}{\gamma_2} \\ P_1 f'_1(N) & f_1(N) - m_1 - 2\gamma P_1 P_2^2 & -2\gamma P_1^2 P_2 \\ P_2 f'_2(N) & 0 & f_2(N) - m_2 \end{pmatrix}.$$

At  $E_0 = (N_0, 0, 0)$  we have

$$V(N_0, 0, 0) = \begin{pmatrix} -D & -\frac{f_1(N_0)}{\gamma_1} & -\frac{f_2(N_0)}{\gamma_2} \\ 0 & f_1(N_0) - m_1 & 0 \\ 0 & 0 & f_2(N_0) - m_2 \end{pmatrix}$$

with eigenvalues  $\alpha_1 = -D$ ,  $\alpha_2 = f_1(N_0) - m_1$ , and  $\alpha_3 = f_2(N_0) - m_2$ . Thus  $E_0$  is locally asymptotically stable for (1) provided  $f_1(N_0) < m_1$  (so that  $N_0 < \lambda_1$ ) and  $f_2(N_0) < m_2$  (so that  $N_0 < \lambda_2$ ): that is, no other equilibria exist. It is unstable if either  $N_0 > \lambda_1$  (so that  $E_{\lambda_1}$  exists) or  $N_0 > \lambda_2$  (so that  $E_{\lambda_2}$  exists).

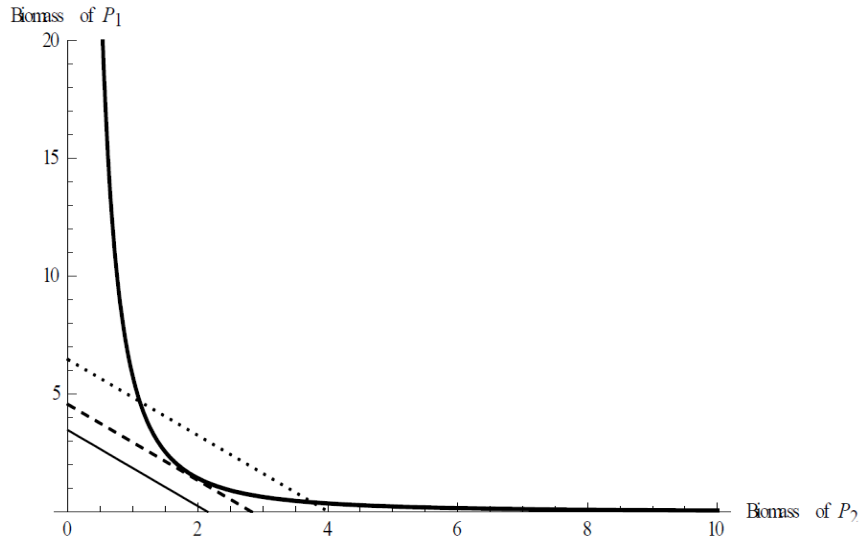


FIGURE 1. Existence and non-existence of interior equilibria of model (1) in the  $(P_1, P_2)$ -plane. For  $N_0 < N_0^c$  there is no interior equilibrium (the graph of the first equation of system (5) does not intersect the curve). For  $N_0 = N_0^c$  there is a unique interior equilibrium (the graph of the first equation of system (5) is tangent to the curve). For  $N_0 > N_0^c$  there are precisely two interior equilibria (the graph of the first equation of system (5) intersects the curve in two distinct points).

Suppose  $N_0 > \lambda_1$  so that  $E_{\lambda_1} = (\lambda_1, \bar{P}_1, 0)$  exists. We examine the local stability properties of  $E_{\lambda_1}$ . At  $E_{\lambda_1}$  the variational matrix is given by

$$V(\lambda_1, \bar{P}_1, 0) = \begin{pmatrix} -D - \frac{\bar{P}_1}{\gamma_1} f_1'(\lambda_1) & -\frac{m_1}{\gamma_1} & -\frac{f_2(\lambda_1)}{\gamma_2} \\ \bar{P}_1 f_1'(\lambda_1) & 0 & 0 \\ 0 & 0 & f_2(\lambda_1) - m_2 \end{pmatrix}.$$

The characteristic polynomial is given by

$$[(f_2(\lambda_1) - m_2) - \alpha] \left[ \alpha^2 + \alpha \left( D + \frac{\bar{P}_1}{\gamma_1} \right) f_1'(\lambda_1) + \frac{m_1 \bar{P}_1}{\gamma_1} f_1'(\lambda_1) \right].$$

Since  $f_1'(N) > 0$  for all  $N \geq 0$ , and  $\bar{P}_1 \geq 0$ , the Routh-Hurwitz criterion implies that the roots of  $\alpha^2 + (D + \frac{\bar{P}_1}{\gamma_1}) f_1'(\lambda_1) \alpha + \frac{m_1 \bar{P}_1}{\gamma_1} f_1'(\lambda_1)$  have negative real parts. Hence,  $E_{\lambda_1}$  is locally asymptotically stable for (1) provided  $f_2(\lambda_1) < m_2$  (so that  $\lambda_1 < \lambda_2$ ) and is unstable if  $f_2(\lambda_1) > m_2$  (so that  $\lambda_1 > \lambda_2$ ).

Assuming that  $N_0 > \lambda_2$ , so that  $E_{\lambda_2} = (\lambda_2, 0, \bar{P}_2)$  exists. The variational matrix at  $E_{\lambda_2}$  is given by

$$V(\lambda_2, 0, \bar{P}_2) = \begin{pmatrix} -D - \frac{\bar{P}_2}{\gamma_1} f_2'(\lambda_2) & -\frac{f_1(\lambda_2)}{\gamma_1} & -\frac{m_2}{\gamma_2} \\ 0 & f_1(\lambda_2) - m_1 & 0 \\ \bar{P}_2 f_2'(\lambda_2) & 0 & 0 \end{pmatrix}.$$

The symmetrical property of  $V(\lambda_2, 0, \bar{P}_2)$  and  $V(\lambda_1, \bar{P}_1, 0)$  allows us to draw the following conclusion:  $E_{\lambda_2}$  is locally asymptotically stable for (1) if and only if  $f_1(\lambda_2) < m_1$  (so that  $\lambda_2 < \lambda_1$ ) and is unstable if  $f_1(\lambda_2) > m_1$  (so that  $\lambda_2 > \lambda_1$ ).

Now suppose that  $N_0 \geq N_0^c$  so that there exists at least one interior equilibrium of the form  $\hat{E} = (\lambda_2, \hat{P}_1, \hat{P}_2)$ . The variational matrix  $V(\lambda_2, P_1, P_2)$  of (1) at an equilibrium point of the form  $(\lambda_2, P_1, P_2)$  (where both  $P_1$  and  $P_2$  are positive) is given by

$$\begin{pmatrix} -D - \frac{P_1 f_1'(\lambda_2)}{\gamma_1} - \frac{P_2 f_2'(\lambda_2)}{\gamma_2} & -\frac{f_1(\lambda_2)}{\gamma_1} & -\frac{m_2}{\gamma_2} \\ P_1 f_1'(\lambda_2) & -m_1 + f_1(\lambda_2) - 2\gamma P_2^2 P_1 & -2\gamma P_1^2 P_2 \\ P_2 f_2'(\lambda_2) & 0 & 0 \end{pmatrix}.$$

The corresponding characteristic polynomial is given by  $r(\alpha) = \alpha^3 + A\alpha^2 + B\alpha + C$ , where

$$\begin{aligned} A &= D + \frac{P_1 f_1'(\lambda_2)}{\gamma_1} + \frac{P_2 f_2'(\lambda_2)}{\gamma_2} + \gamma P_2^2 P_1 > 0 \\ B &= \frac{P_2 f_2'(\lambda_2) m_2}{\gamma_2} + \frac{P_1 f_1'(\lambda_2) f_1(\lambda_2)}{\gamma_1} + \gamma \left( D + \frac{P_1 f_1'(\lambda_2)}{\gamma_1} + \frac{P_2 f_2'(\lambda_2)}{\gamma_2} \right) P_2^2 P_1 > 0 \\ C &= \gamma P_2^2 P_1 f_2'(\lambda_2) \left[ \frac{m_2 P_2}{\gamma_2} - \frac{2P_1 f_1(\lambda_2)}{\gamma_1} \right]. \end{aligned} \tag{7}$$

We have three cases to consider for the different interior equilibrium points.

**Case 1.** At the critical point  $\tilde{E} = (\lambda_2, \tilde{P}_1, \tilde{P}_2)$ , where  $\tilde{P}_1 = \frac{\gamma_1(N_0 - \lambda_2)D}{3f_1(\lambda_2)}$  and  $\tilde{P}_2 = \frac{2\gamma_2(N_0 - \lambda_2)D}{3m_2}$ , replacing each  $P_i$  in the expression for  $C$  with  $\tilde{P}_i$  yields  $C = 0$ . Thus, zero is an eigenvalue of  $V(\lambda_2, \tilde{P}_1, \tilde{P}_2)$ , so that  $\tilde{E} = (\lambda_2, \tilde{P}_1, \tilde{P}_2)$  is a nonhyperbolic rest point of model (1).

In the remaining cases, we use the following notations for the interior equilibria when  $N_0 > N_0^c$ :  $\hat{E}' = (\lambda_2, \hat{P}'_1, \hat{P}'_2)$  and  $\hat{E}'' = (\lambda_2, \hat{P}''_1, \hat{P}''_2)$  with  $0 < \hat{P}'_1 < \tilde{P}_1$ ,  $\hat{P}'_2 < \tilde{P}'_2 < 3\tilde{P}_2/2$ ,  $\hat{P}'_1 < \hat{P}''_1 < 3\tilde{P}_1$ , and  $0 < \hat{P}''_2 < \tilde{P}_2$ .

**Case 2.** At the critical point  $\hat{E}' = (\lambda_2, \hat{P}'_1, \hat{P}'_2)$ , a direct computation of  $C$  leads to  $C > 0$ . By the Routh-Hurwitz criterion, the eigenvalues of  $V(\lambda_2, \hat{P}'_1, \hat{P}'_2)$  have negative real parts if and only if  $A > 0$  (true by definition of  $A$ ) and  $AB > C$ . Hence, when  $\hat{E}'$  exists, it is locally asymptotically stable for (1) if and only if  $AB > C$ .

**Case 3.** At the rest point  $\hat{E}'' = (\lambda_2, \hat{P}''_1, \hat{P}''_2)$ , a direct computation of  $C$  gives us  $C < 0$ . By the Routh-Hurwitz criterion,  $V(\lambda_2, \hat{P}''_1, \hat{P}''_2)$  has at least one eigenvalue with positive real, so that  $\hat{E}''$  is unstable (saddle point).

Roy [38] asserts that, when  $N_0 > N_0^c$  and the functions  $f_i$  are Michaelis-Menten one of these two coexistence equilibria is locally stable whereas the other is unstable. Our results are consistent with those of Roy [38] in that  $\hat{E}'' = (\lambda_2, \hat{P}''_1, \hat{P}''_2)$  is unstable and we are able to give conditions under which  $\hat{E}' = (\lambda_2, \hat{P}'_1, \hat{P}'_2)$  is locally asymptotically stable.

We summarize the results of this subsection in the following theorem.

**Theorem 2.1.** 1.  $E_0$  always exists. It is locally asymptotically stable for (1), if and only if  $N_0 < \lambda_i$  for  $i = 1, 2$ .  
2. Suppose  $\lambda_1 < N_0$ , so that  $E_{\lambda_1} = (\lambda_1, \bar{P}_1, 0)$  exists.  $E_{\lambda_1}$  is locally asymptotically stable for (1) if and only if  $\lambda_1 < \lambda_2$  (so that  $f_2(\lambda_1) < m_2$ ).



3. Suppose  $\lambda_2 < N_0$ , so that  $E_{\lambda_2} = (\lambda_2, 0, \bar{P}_2)$  exists.  $E_{\lambda_2}$  is locally asymptotically stable for (1) if and only if  $\lambda_2 < \lambda_1$  (so that  $f_1(\lambda_2) < m_1$ ).
4. Suppose  $N_0 = N_0^c$ , so that  $\tilde{E} = (\lambda_2, \tilde{P}_1, \tilde{P}_2)$  exists. Then  $\tilde{E}$  is nonhyperbolic.
5. Suppose  $N_0 > N_0^c$ , so that there are two distinct interior equilibria,  $\hat{E}' = (\lambda_2, \hat{P}'_1, \hat{P}'_2)$  and  $\hat{E}'' = (\lambda_2, \hat{P}''_1, \hat{P}''_2)$  with  $0 < \hat{P}'_1 < \tilde{P}_1$ ,  $\tilde{P}_2 < \hat{P}'_2 < 3\tilde{P}_2/2$ ,  $\tilde{P}_1 < \hat{P}''_1 < 3\tilde{P}_1$ , and  $0 < \hat{P}''_2 < \tilde{P}_2$ . Then  $\hat{E}''$  is a saddle, while  $\hat{E}' = (\lambda_2, \hat{P}'_1, \hat{P}'_2)$  satisfying (8), is locally asymptotically stable for (1) if and only if  $AB > C$ .

$$\begin{aligned}
 A &= D + \frac{\hat{P}'_1 f'_1(\lambda_2)}{\gamma_1} + \frac{\hat{P}'_2 f'_2(\lambda_2)}{\gamma_2} + \gamma \hat{P}'_2{}^2 \hat{P}'_1 > 0 \\
 B &= \frac{\hat{P}'_2 f'_2(\lambda_2) m_2}{\gamma_2} + \frac{\hat{P}'_1 f'_1(\lambda_2) f_1(\lambda_2)}{\gamma_1} + \gamma \left( D + \frac{\hat{P}'_1 f'_1(\lambda_2)}{\gamma_1} + \frac{\hat{P}'_2 f'_2(\lambda_2)}{\gamma_2} \right) \hat{P}'_2{}^2 \hat{P}'_1 > 0 \\
 C &= \gamma \hat{P}'_2{}^2 \hat{P}'_1 f'_2(\lambda_2) \left[ \frac{m_2 \hat{P}'_2}{\gamma_2} - \frac{2\hat{P}'_1 f_1(\lambda_2)}{\gamma_1} \right].
 \end{aligned}
 \tag{8}$$

**2.3. Ecological interpretations of inequalities.** This subsection gives ecological interpretations of inequalities resulting from the local stability results for our model (1) (see Theorem 2.1). By statement 1 of Theorem 2.1, the species-free steady state is locally asymptotically stable if and only if  $f_i(N_0) < m_i$ , for  $i = 1, 2$ . That is, the growth rate of species  $P_i$  is strictly less than its removal rate  $m_i$ , even when the culture vessel is held at the input nutrient concentration  $N_0$ . Thus neither  $P_1$  nor  $P_2$  can survive at this input level of nutrient.

Statement 2 means that the growth rate of species  $P_2$  is strictly less than its removal rate when the nutrient level in the culture vessel is held at  $\lambda_1$ . Thus, species  $P_2$  cannot compensate for the rate at which it is being removed from competition. It makes biological sense that only  $P_1$  avoids extinction for initial conditions in a neighborhood of  $E_{\lambda_1}$ .

The biological interpretation of the local stability conditions for  $E_{\lambda_2}$  is symmetrical to that of  $E_{\lambda_1}$ .

**2.4. Global results.** In this subsection, we establish the global asymptotical stability of boundary equilibria due to the inadequacy of the resource supply. We first establish the competition-independent extinction of  $P_i$ . The proof uses the following result due to Miller [28].

**Lemma 2.2.** (Miller’s Lemma) Let  $\omega(t) \in C^2(t_0, \infty)$ ,  $\omega(t) \geq 0$  and  $K > 0$ .

(a) If  $\omega'(t) \geq 0$ ,  $\omega(t)$  is bounded and  $\omega''(t) \leq K$  for all  $t \geq t_0$  then  $\omega'(t) \rightarrow 0$  as  $t \rightarrow \infty$ .

(b) If  $\omega'(t) \leq 0$ ,  $\omega''(t) \leq -K$  for all  $t \geq t_0$  then  $\omega'(t) \rightarrow 0$  as  $t \rightarrow \infty$ .

**Lemma 2.3.** If  $\lambda_i > N_0$ ; then  $P_i(t) \rightarrow 0$  as  $t \rightarrow \infty$  in (1).

*Proof of Lemma 2.3.* Choose  $\delta > 0$  so that  $N_0 + \delta < \lambda_i$ . By Lemma 2.1(b),  $N(t) < N_0 + \delta$  for all sufficiently large  $t$ . From the second and third equations of system (1), and by monotonicity properties of uptake functions  $f_i$ , we have

$$\frac{dP_i(t)}{dt} \leq P_i(t)[f_i(N(t)) - m_i] \leq P_i(t)[f_i(N_0 + \delta) - m_i]
 \tag{9}$$

for all sufficiently large  $t$ . Hence by the definition of  $\lambda_i$  (that is,  $f_i(\lambda_i) = m_i$ ),  $\frac{dP_i(t)}{dt} < 0$  for all sufficiently large  $t$ . Also,  $P_i''(t)$  is bounded below. It follows

from Lemma 2.2 (b) that  $P_i'(t) \rightarrow 0$  as  $t \rightarrow 0$ . However,  $\limsup_{t \rightarrow \infty} f_i(N(t)) < f_i(N_0 + \delta) < m_i$  so that the only possibility is that  $P_i(t) \rightarrow 0$  as  $t \rightarrow \infty$ .  $\square$

We are now in a position to prove that  $E_0$  is a global attractor when it is the only steady state.

**Theorem 2.2.** *If  $N_0 < \lambda_i$  for  $i = 1, 2$ ,  $E_0$  is globally asymptotically stable for (1).*

*Proof.* Take  $Q \in \{(N, P_1, P_2) \in \mathbf{R}_+^3 : P_1 > 0, P_2 > 0\}$ . Let  $\Omega(Q)$  denote the omega limit set of the orbit through  $Q$ . By the hypothesis and Lemma 2.3, any  $P = (N, P_1, P_2) \in \Omega(Q)$  satisfies  $P_1 = 0$  and  $P_2 = 0$ . On  $\{(N, 0, 0) \in \mathbf{R}_+^3\}$  the system reduces to  $N'(t) = (N_0 - N(t))D$  and hence  $N(t) \rightarrow N_0$ . Therefore,  $\{E_0\} \in \Omega(Q)$ . Since all solutions of (1) are positive and bounded,  $\Omega(Q)$  is a nonempty and compact subset of  $\mathbf{R}_+^3$ . If  $P \in \Omega(Q)$  then the entire trajectory through  $P$  is in  $\Omega(Q)$ . Hence,  $E_0$  is the only candidate. Thus,  $E_0$  is globally asymptotically stable for (1).  $\square$

The next theorem gives conditions under which  $E_{\lambda_i}$  is globally asymptotically stable.

**Theorem 2.3.** (a) *Suppose  $\lambda_1 < N_0 < \lambda_2$ . Then  $E_{\lambda_1}$  is globally asymptotically stable for (1).*

(b) *Suppose  $\lambda_2 < N_0 < \lambda_1$ . Then  $E_{\lambda_2}$  is globally asymptotically stable for (1).*

*Proof.* Here we need to prove only (a) as the proof of (b) is similar. Take

$$Q \in \{(N, P_1, P_2) \in \mathbf{R}_+^3 : P_1 > 0, P_2 > 0\}.$$

Let  $\Omega(Q)$  denote the omega limit set of the orbit through  $Q$ . By the hypothesis and Lemma 2.3, any  $P = (N, P_1, P_2) \in \Omega(Q)$  satisfies  $P_2 = 0$ . On  $\{(N, P_1, 0) \in \mathbf{R}_+^3\}$  the system reduces to

$$\begin{aligned} N'(t) &= (N_0 - N(t))D - \frac{P_1(t)}{\gamma_1} f_1(N(t)) \\ P_1'(t) &= P_1(t)(f_1(N(t)) - m_1). \end{aligned} \tag{10}$$

By an argument comparable to that given in Hsu [21],  $N(t) \rightarrow \lambda_1$  and  $P_1(t) \rightarrow \bar{P}_1 = \frac{\gamma_1(N_0 - \lambda_1)D}{m_1}$ . Therefore,  $\{E_{\lambda_1}\} \in \Omega(Q)$ . Since (10) has no periodic orbits and the boundary is acyclic, it follows from Lemma 4.3 in Thieme [45] that  $E_{\lambda_1}$  is globally asymptotically stable for (1).  $\square$

**2.5. Transfer of local stability and one-parameter bifurcation.** In this subsection we describe the evolution of equilibria into the nonnegative cone of  $\mathbf{R}^3$  and the consequent transfer of stability as  $N_0$  is increased. In particular, whenever a new boundary steady state coalesces with an existing one, a transcritical bifurcation occurs. We have two scenarios to consider.

Scenario 1:  $\lambda_1 < \lambda_2$ . When  $N_0 < \lambda_1$  only the washout equilibrium  $E_0$  exists and is globally asymptotically stable. As  $N_0$  is increased so that  $N_0 = \lambda_1$ ,  $E_0$  and  $E_{\lambda_1}$  coalesce. As  $N_0$  is increased still further, so that  $\lambda_1 < N_0 < \lambda_2$ ,  $E_{\lambda_1}$  bifurcates into the nonnegative cone of  $\mathbf{R}^3$  through  $E_0$ . The washout equilibrium loses a degree of stability, and  $E_{\lambda_1}$  is globally asymptotically stable for (1). As  $N_0$  is increased so that  $N_0 = \lambda_2$ ,  $E_0$  and  $E_{\lambda_2}$  coalesce. As  $N_0$  is increased still further so that  $\lambda_1 < \lambda_2 < N_0 < N_0^c$ ,  $E_{\lambda_2}$  bifurcates into the nonnegative cone of  $\mathbf{R}^3$  through  $E_0$ . Here,  $E_0$  loses another degree of stability,  $E_{\lambda_1}$  loses a degree of stability, and  $E_{\lambda_2}$  is locally asymptotically stable for (1). When  $N_0 = N_0^c$  the interior equilibrium  $\bar{E}$

becomes feasible in the positive cone of  $\mathbf{R}^3$ . For  $N_0 > N_0^c$ ,  $\tilde{E}$  undergoes a saddle-node bifurcation to generate  $\hat{E}'$  and  $\hat{E}''$ .  $\hat{E}''$  is a saddle point for model (1) while  $\hat{E}'$  is locally asymptotically stable for (1) under conditions given in Theorem 2.1. In other words, we have an unstable interior equilibrium and an asymptotically stable interior equilibrium. As such, persistence is not possible. Unlike Roy [38], our goal is not to identify the basin of attraction of the locally stable interior equilibrium. We are more interested in global stability and uniform persistence.

Scenario 2:  $\lambda_2 < \lambda_1$ . When  $N_0 < \lambda_2$  only the washout equilibrium  $E_0$  exists and is globally asymptotically stable. As  $N_0$  is increased so that  $N_0 = \lambda_2$ ,  $E_0$  and  $E_{\lambda_2}$  coalesce. As  $N_0$  is increased still further, so that  $\lambda_2 < N_0 < \lambda_1$ ,  $E_{\lambda_2}$  bifurcates into the nonnegative cone of  $\mathbf{R}^3$  through  $E_0$ . The washout equilibrium loses a degree of stability, and  $E_{\lambda_2}$  is globally asymptotically stable. As  $N_0$  is increased so that  $N_0 = \lambda_1$ ,  $E_0$  and  $E_{\lambda_1}$  coalesce. As  $N_0$  is increased still further so that  $\lambda_2 < \lambda_1 < N_0$ ,  $E_{\lambda_1}$  bifurcates into the nonnegative cone of  $\mathbf{R}^3$  through  $E_0$ . Here,  $E_0$  loses another degree of stability,  $E_{\lambda_2}$  loses a degree of stability, and  $E_{\lambda_1}$  is locally asymptotically stable for (1), attracting solutions from the interior.

Hence, system (1) cannot be uniformly persistent.

**2.6. Effect of nutrient recycling.** The model considered so far lacks the effects of nutrient recycling. The regeneration of nutrient due to bacterial decomposition of the dead biomass must be considered. We modify model (1) to incorporate the effect of nutrient recycled from each phytoplankton cell on its death. The mortality of nontoxic phytoplankton  $P_1$  is the sum of its intrinsic death and the death due to phytotoxic effects. The mortality of toxic phytoplankton  $P_2$  is due to its intrinsic death and the death due autotoxic effects. For simplicity, we assume limiting nutrient recycling is instantaneous. We denote by  $\eta_i$  (assumed to be less than 1 and constant over time) the nutrient contents of a single cell of phytoplankton  $P_i$ . Under these considerations, model (1) can be extended as follows:

$$\begin{aligned} \frac{dN}{dt} &= (N_0 - N)D - \frac{1}{\gamma_1} f_1(N)P_1 - \frac{1}{\gamma_2} f_2(N)P_2 \\ &\quad + \eta_1(\epsilon_1 + \gamma P_1 P_2^2)P_1 + \eta_2(\epsilon_2 + a_2)P_2, \\ \frac{dP_1}{dt} &= f_1(N)P_1 - m_1 P_1 - \gamma P_1^2 P_2^2 \\ \frac{dP_2}{dt} &= f_2(N)P_2 - m_2 P_2, \\ N(0) &> 0, P_1(0) > 0, P_2(0). \end{aligned} \tag{11}$$

We assume that all model parameters are nonnegative. Adapting the method used for the previous model, we can show that model (11) has the following steady states:  $E_0^* = (N_0, 0, 0)$ ,  $E_{\lambda_1}^* = (\lambda_1, \frac{\gamma_1(N_0 - \lambda_1)D}{m_1 - \gamma_1 \eta_1 \epsilon_1}, 0)$  with  $N_0 > \lambda_2$  and  $m_1 > \gamma_1 \eta_1 \epsilon_1$ , and  $E_{\lambda_2}^* = (\lambda_2, 0, \frac{\gamma_2(N_0 - \lambda_2)D}{m_2 - \gamma_2 \eta_2 (\epsilon_2 + a_2)})$  with  $N_0 > \lambda_2$  and  $m_2 > \gamma_2 \eta_2 (\epsilon_2 + a_2)$ .

As before, there are zero, one or two interior equilibria depending on whether or not the magnitude of the input nutrient concentration equals or exceeds a modified critical value  $N_0^{c*}$  given by,

$$N_0^{c*} = \lambda_2 + \sqrt[3]{\frac{27(f_1(\lambda_2) - \gamma_1 \eta_1 (f_1(\lambda_2) - D))(m_2 - \gamma_2 \eta_2 (\epsilon_2 + a_2))^2 (f_1(\lambda_2) - m_1)}{4\gamma_1 \gamma_2^2 D^3 \gamma}}. \tag{12}$$

The qualitative dynamics remain unchanged from those exhibited by system (1) when the effect of nutrient recycling is incorporated into the model. Only the critical values of the parameters at which transitions take place are affected. For instance, the eigenvalues of the Jacobian matrix at  $E_0^*$  are  $\alpha_1^* = -D$ ,  $\alpha_2^* = f_1(N_0) - m_1$ , and  $\alpha_3^* = f_2(N_0) - m_2$ . Thus the local stability of  $E_0$  remains the same under nutrient recycling. The characteristic polynomial of the Jacobian matrix at  $E_{\lambda_1}^* = (\lambda_1, \frac{\gamma_1(N_0 - \lambda_1)D}{m_1 - \gamma_1\eta_1\epsilon_1}, 0)$  is given by

$$[(f_2(\lambda_1) - m_2) - \alpha] \left[ \alpha^2 + \alpha \left( D + \frac{\bar{P}_1}{\psi_1 \gamma_1} \right) f_1'(\lambda_1) + \left( \frac{m_1 \bar{P}_1}{\gamma_1} - \eta_1 \epsilon_1 \right) f_1'(\lambda_1) \right],$$

where  $\psi_1 = 1 - \frac{\gamma_1 \eta_1 \epsilon_1}{m_1}$ . The existence conditions of  $E_{\lambda_1}^*$  and the Routh-Hurwitz criterion imply that the roots of this polynomial have negative real parts. Hence the local stability criteria of  $E_{\lambda_1}^*$  and  $E_{\lambda_1}$  are similar. Observe that along a given  $P_i$ -direction, each equilibrium of model (11) appears below the corresponding equilibrium of model (1).

Our analytical findings are consistent with those of Roy [38]. In the next section, we further extend this work to include predation of phytoplankton species by a zooplankton population.

**3. The competition-allelopathy-predation model.** We reconsider model (1) and add a zooplankton species  $Z$  as a fourth interacting component. The two phytoplankton populations are the only prey of zooplankton. We neglect the potential negative effect of toxic phytoplankton  $P_2$  on the growth of zooplankton  $Z$ . With the above assumptions our model may be formulated as follows:

$$\begin{aligned} \frac{dN}{dt} &= (N_0 - N)D - \frac{1}{\gamma_1} f_1(N)P_1 - \frac{1}{\gamma_2} f_2(N)P_2, \\ \frac{dP_1}{dt} &= f_1(N)P_1 - m_1 P_1 - \frac{1}{\eta_1} g_1(P_1)Z - \gamma P_1^2 P_2^2, \\ \frac{dP_2}{dt} &= f_2(N)P_2 - m_2 P_2 - \frac{1}{\eta_2} g_2(P_2)Z \\ \frac{dZ}{dt} &= (g_1(P_1) + g_2(P_2) - c)Z, \\ N(0) &> 0, P_1(0) > 0, P_2(0) > 0, Z(0) > 0. \end{aligned} \tag{13}$$

In these equations,  $N(t)$ ,  $P_1(t)$ ,  $P_2(t)$ ,  $D$ ,  $\gamma_i$ , and  $m_i$  have the same meanings as in model (1) and the  $f_i$ 's satisfy assumptions of model (1).  $Z(t)$  is the biomass of zooplankton species at time  $t$ . Here, the removal rate  $c$  of zooplankton  $Z$  is the sum of washout rate  $D$  and the specific death rate  $\xi$ , so that  $c = D + \xi$ . The response function  $g_i(P_i)$  represents the per capita rate of conversion of phytoplankton  $P_i$  to biomass of population  $Z$  per unit of population  $Z$  as a function of the biomass of phytoplankton  $P_i$ . We assume that the rate of conversion of biomass of phytoplankton  $P_i$  to biomass of zooplankton  $Z$  is proportional to the amount of phytoplankton consumed, so that the consumption rate of phytoplankton  $P_i$  per unit of population  $Z$  is of the form  $\frac{1}{\eta_i} g_i(P_i)$ , where  $\eta_i$  is a growth yield constant (number of zooplankton per unit of phytoplankton). Since  $P_1$  and  $P_2$  are perfectly substitutable resources for  $Z$  (see for examples, Butler and Wolkowicz [8], León and Tumpson [25], Rapport [35], and Ballyk and Wolkowicz [4]), the per capita growth rate of zooplankton as a function of  $P_1$  and  $P_2$  takes the form  $G(P_1, P_2) = g_1(P_1) + g_2(P_2)$  for all  $P_1 \geq 0$

and  $P_2 \geq 0$ . Following Li and Kuang [26], Ruan [40], Wolkowicz and Lu [48] and others, we make the following assumptions concerning the response functions  $g_i$ :

$$g_i : \mathbf{R}_+ \longrightarrow \mathbf{R}_+, g_i \text{ is continuously differentiable,} \tag{14}$$

$$g_i(0) = 0, g'_i(P_i) > 0, \lim_{P_i \rightarrow \infty} g_i(P_i) = \omega_i, \text{ for all } P_i \geq 0, \tag{15}$$

where  $\omega_i$  denotes the maximal growth rate of zooplankton  $Z$  on phytoplankton  $P_i$ .

It will also be convenient to express  $g_i(P_i)$  as

$$g_i(P_i) = P_i h_i(P_i), \tag{16}$$

where  $h_i(P_i)$  is some positive and differentiable function. Since  $g_i$  is continuously differentiable it follows that

$$\lim_{P_i \rightarrow 0} h_i(P_i) = g'_i(0), \tag{17}$$

and so we define

$$h_i(0) = g'_i(0). \tag{18}$$

The breakeven concentration for population  $Z$  on phytoplankton  $P_i$  is obtained by setting  $\frac{dZ}{dt} = 0 = g_i(P_i) - c$  and solving for  $P_i$ . By the monotonicity assumptions, the solution  $\Lambda_i$  is a uniquely defined positive extended real number as long as we assume  $\Lambda_i = \infty$  if  $g_i(P_i) < c$  for all  $P_i \geq 0$ .

Prototypes of response functions  $g_i$  often found in the literature (Anderson *et al.* [2], Aris and Humphrey [3], Boon and Laudelout [6], Edwards [12], Jost *et al.* [23], Monod [30], Ruan [40], Wolkowicz and Lu [48], Yang and Humphrey [49]) include Lotka-Volterra kinetics, Michaelis-Menten kinetics, sigmoidal kinetics, and Ivlev's functional response formulation.

The half-saturation constant  $L_i$  of zooplankton for the  $i^{th}$  phytoplankton is given by  $g_i(L_i) = \frac{\omega_i}{2}$  and represents the phytoplankton biomass  $P_i$  supporting growth at half the maximal growth rate.

System (13) was considered by Butler and Wolkowicz [7] under the assumptions that specific death rates are insignificant compared to the washout rate  $D$  ( $\epsilon_i = 0, i = 1, 2, \xi = 0$ ),  $Z$  feeds only on one phytoplankton  $P_i$  and there is no allelopathic effect. Holt *et al.* [19] studied model (13) under linearity of response functions  $f_i$  and  $g_i$ , and in the absence of allelopathic effects. Grover and Holt [13] relaxed the linearity assumptions on the responses functions in Holt *et al.* [19] and included the Holling types I and II response functions. In addition, system (13) was considered by Li and Kuang [26] under the assumption that one of the  $P_i$ 's is absent and there is no allelopathic interaction. As such, model (13) is a significant generalization of those previously considered.

**3.1. Preliminary results.** We first establish that solutions of (13) are positive and bounded.

**Lemma 3.1.** *Solutions of (13) are (a) positive and (b) bounded.*

*Proof of (a).* The proof is similar to that of Lemma 2.1 (a) except that we have to replace the  $A_i$ 's and  $t_i$ 's in the proofs of  $P_i(t) > 0$  for all  $t$ , by the following  $\hat{A}_i$ 's and  $\hat{t}_i$ 's, respectively:

For  $i \in 1, 2$ , let  $\hat{t}_i = \min \{t > 0 : P_i(t) = 0\}$  and define

$$\hat{A}_1 = \min \left\{ f_1(N(t)) - m_1 - \frac{1}{\eta_1} h_1(P_1(t))Z(t) - \gamma P_1(t)P_2^2(t), 0 \leq t \leq \hat{t}_1 \right\}, \quad (19a)$$

$$\hat{A}_2 = \min \left\{ f_2(N(t)) - m_2 - \frac{1}{\eta_2} h_2(P_2(t))Z(t), 0 \leq t \leq \hat{t}_2 \right\}. \quad (19b)$$

Suppose now that  $Z(0) > 0$ , then  $\frac{dZ}{dt} = (g_1(P_1) + g_2(P_2))Z - cZ$  yields  $Z(t) = Z(0) \exp\left(\int_0^t h(s) ds\right) > 0$ , where  $h(s) = g_1(P_1(s)) + g_2(P_2(s)) - c$ .  $\square$

*Proof of (b).* To prove boundedness of solutions of model (13), define  $T(t) = N(t) + \frac{1}{\gamma_1}P_1(t) + \frac{1}{\gamma_2}P_2(t) + Z(t)$ . From (13) we have

$$\frac{dT}{dt} \leq N_0D - D_0T,$$

where  $D_0 = \min(D, m_1, m_2, c)$ . By Gronwall's lemma,  $0 \leq T(t) \leq \frac{N_0D}{D_0} + (T(0) - \frac{N_0D}{D_0}) \exp(-D_0t)$ . Since  $\lim_{t \rightarrow \infty} (T(0) - \frac{N_0D}{D_0}) \exp(-D_0t) = 0$ , it follows that for all  $\epsilon > 0$ , the solutions  $(N(t), P_1(t), P_2(t), Z(t))$  of (13) satisfy

$$0 \leq N(t) + \frac{1}{\gamma_1}P_1(t) + \frac{1}{\gamma_2}P_2(t) + Z(t) \leq \frac{N_0D}{D_0} + \epsilon, \quad (20)$$

for sufficiently large  $t$ .  $\square$

The Fundamental Existence-Uniqueness Theorem (see, for example, Perko [33]) and Lemma 3.1(b) ensure that solutions of (13) exist uniquely for all time.

**3.2. Steady states: Existence and local stability.** It is straightforward to prove that all the boundary steady states of model (1) along with their existence conditions are transferred to model (13) except that we add a zero  $Z$  component. In the remainder of this work, we denote by  $\lambda_i$  and  $\Lambda_i$  the unique positive solutions of  $f_i(N) = m_i$ , and  $g_i(P_i) = c$ , respectively. The steady states of model (13) along with their existence conditions are listed below.

The washout equilibrium  $E_0 = (N_0, 0, 0, 0)$  always exists.  $E_{\lambda_1} = (\lambda_1, \bar{P}_1, 0, 0)$  exists provided  $N_0 > \lambda_1$ .  $E_{\lambda_2} = (\lambda_2, 0, \bar{P}_2, 0)$  exists as long as  $N_0 > \lambda_2$ . For simplicity, we have kept the same notations for  $E_0$ ,  $E_{\lambda_1}$ , and  $E_{\lambda_2}$  as in model (1).

No equilibrium of the form  $(N, 0, 0, Z)$  (for which  $N$  and  $Z$  are both positive) exists.  $E_{\Lambda_1} = (N_1, \Lambda_1, 0, Z_1)$  exists if and only if  $\lambda_1 < N_1 < N_0 - \frac{\Lambda_1 m_1}{\gamma_1 D}$ . Similarly,  $E_{\Lambda_2} = (N_2, 0, \Lambda_2, Z_2)$  exists provided  $\lambda_2 < N_2 < N_0 - \frac{\Lambda_2 m_2}{\gamma_2 D}$ . Here  $N_i$  is the unique positive solution of  $\gamma_i(N_0 - N)D = \Lambda_i f_i(N)$  when  $\Lambda_i$ , while  $Z_i = \frac{\eta_i(f_i(N_i) - m_i)}{h_i(\Lambda_i)}$ , for  $i = 1, 2$ .

The zooplankton-free equilibria (for which for which  $N, P_1, P_2$  are all positive and  $Z = 0$ ) depend on the magnitude  $N_0$  and are derived from the interior equilibria of model (1) by just adding a zero  $Z$  component. As established in Subsection 2.2, there is a critical value of  $N_0$

$$N_0^c = \lambda_2 + \sqrt[3]{\frac{27f_1(\lambda_2)m_2^2(f_1(\lambda_2) - m_1)}{4\gamma_1\gamma_2^2D^3\gamma}}. \quad (21)$$

Model (13) does not have any equilibrium point of the form  $(N, P_1, P_2, 0)$  (where  $N, P_1$ , and  $P_2$  are all positive) when  $N_0 < N_0^c$ . When  $N_0 = N_0^c$ , model (13) has precisely one equilibrium given by  $\tilde{E} = (\lambda_2, \tilde{P}_1, \tilde{P}_2, 0)$ , where  $\tilde{P}_1 = \gamma_1(N_0 -$

$\lambda_2)D/3f_1(\lambda_2)$ , and  $\tilde{P}_2 = 2\gamma_2(N_0 - \lambda_2)D/3m_2$ . As  $N_0$  is increased beyond  $N_0^c$ , the straight line intersects the curve in two points and (13) has two equilibria of the form  $\bar{E}' = (\lambda_2, \bar{P}_1', \bar{P}_2', 0)$  and  $\bar{E}'' = (\lambda_2, \bar{P}_1'', \bar{P}_2'', 0)$ , where  $0 < \bar{P}_1' < \tilde{P}_1, \tilde{P}_2 < \bar{P}_2' < 3\tilde{P}_2/2, \tilde{P}_1 < \bar{P}_1'' < 3\tilde{P}_1$ , and  $0 < \bar{P}_2'' < \tilde{P}_2$  (see Figure 1).

The local stability properties of (13) through an examination of the linearized system about the equilibria  $E_0, E_{\lambda_1}$ , and  $E_{\lambda_2}$  are omitted because they are similar to those of the corresponding boundary equilibria in Theorem 2.1.

Let  $\hat{E} = (\lambda_2, \hat{P}_1, \hat{P}_2, 0)$  be an arbitrary zooplankton-free critical point for (13). The stability conditions for  $\bar{E}'$  are similar to those of an interior equilibrium for (1) except that we must add the inequality

$$g_1(\bar{P}_1') + g_2(\bar{P}_2') < c.$$

We will just investigate the local stability properties of (13) through an examination of the linearized system about the equilibria  $E_{\Lambda_1}$ , and  $E_{\Lambda_2}$ .

We assume that  $N_0 - \frac{\Lambda_2 m_2}{\gamma_2 D} > N_2 > \lambda_2$  and  $f_2(N_2) > m_2$ , so that  $E_{\Lambda_2}$  exists. We examine the local stability properties of  $E_{\Lambda_2}$ . The variational matrix of (13) evaluated at  $E_{\Lambda_2}$ , is given by

$$\begin{pmatrix} -D - \frac{\Lambda_2 f_2'(N_2)}{\gamma_2} & -\frac{f_1(N_2)}{\gamma_1} & -\frac{f_2(N_2)}{\gamma_2} & 0 \\ 0 & f_1(N_2) - m_1 - \frac{Z_2}{\eta_1} g_1'(0) & 0 & 0 \\ \Lambda_2 f_2'(N_2) & 0 & f_2(N_2) - m_2 - \frac{Z_2}{\eta_2} g_2'(\Lambda_2) & -\frac{c}{\eta_2} \\ 0 & Z_2 g_1'(0) & Z_2 g_2'(\Lambda_2) & 0 \end{pmatrix}.$$

The corresponding characteristic polynomial is given by

$$p(\alpha) = (\alpha - f_1(N_2) + m_1 + \frac{Z_2}{\eta_1} g_1'(0))(\alpha^3 + \tilde{A}\alpha^2 + \tilde{B}\alpha + \tilde{C}),$$

where

$$\tilde{A} = -f_2(N_2) + m_2 + \frac{Z_2}{\eta_2} g_2'(\Lambda_2) + D + \frac{\Lambda_2 f_2'(N_2)}{\gamma_2}, \tag{22a}$$

$$\tilde{B} = (D + \frac{\Lambda_2 f_2'(N_2)}{\gamma_2})(-f_2(N_2) + m_2 + \frac{Z_2}{\eta_2} g_2'(\Lambda_2)) + g_2'(\Lambda_2) Z_2 \frac{c}{\eta_2} + \Lambda_2 f_2'(N_2) \frac{f_2(N_2)}{\eta_2}, \tag{22b}$$

$$\tilde{C} = Z_2 g_2'(\Lambda_2) \frac{c}{\eta_2} (D + \frac{\Lambda_2 f_2'(N_2)}{\gamma_2}) > 0. \tag{22c}$$

The monotonicity of  $f_2(N)$  and  $g_2(P_2)$ , the positivity of  $N_2, \Lambda_2$ , and  $Z_2$ , together with the Routh-Hurwitz criterion, ensure that the roots of the cubic factor have negative real parts if and only if  $\tilde{A} > 0$  and  $\tilde{A}\tilde{B} > \tilde{C}$ . Hence,  $E_{\Lambda_2}$  is locally asymptotically stable for (13) if and only if  $f_1(N_2) < \frac{Z_2}{\eta_1} g_1'(0) + m_1, \tilde{A} > 0$  and  $\tilde{A}\tilde{B} > \tilde{C}$ .

The local stability analysis of  $E_{\Lambda_1} = (N_1, \Lambda_1, 0, Z_1)$  is symmetrical to the analysis for  $E_{\Lambda_2}$ . It is straightforward to show that the coefficients of the cubic factor of the corresponding characteristic polynomial are given by

$$\hat{A} = -f_1(N_1) + m_1 + \frac{Z_1}{\eta_1} g_1'(\Lambda_1) + D + \frac{\Lambda_1 f_1'(N_1)}{\gamma_1}, \tag{23a}$$

$$\hat{B} = (D + \frac{\Lambda_1 f'_1(N_1)}{\gamma_1})(-f_1(N_1) + m_1 + \frac{Z_1}{\eta_1} g'_1(\Lambda_1)) + g'_1(\Lambda_1) Z_1 \frac{c}{\eta_1} + \Lambda_1 f'_1(N_1) \frac{f_1(N_1)}{\eta_1}, \tag{23b}$$

$$\hat{C} = Z_1 g'_1(\Lambda_1) \frac{c}{\eta_1} (D + \frac{\Lambda_1 f'_1(N_1)}{\gamma_1}) > 0. \tag{23c}$$

We summarize the results of this subsection in the following theorem which extends Theorem 2.1.

- Theorem 3.1.** 1.  $\hat{E}_0$  is locally asymptotically stable for (13), if and only if  $N_0 < \lambda_i$  for  $i = 1, 2$ .
2. Suppose  $\lambda_1 < N_0$ , so that  $E_{\lambda_1} = (\lambda_1, \bar{P}_1, 0, 0)$  exists.  $E_{\lambda_1}$  is locally asymptotically stable for (13) if and only if  $\lambda_1 < \lambda_2$  (so that  $f_2(\lambda_1) < m_2$ ) and  $g_1(\bar{P}_1) < c$ . The local stability conditions for  $E_{\lambda_2}$  are symmetrical to those of  $E_{\lambda_1}$  and therefore omitted.
3. Assume that  $N_0 - \frac{\Lambda_1 m_1}{\gamma_1 D} > N_1 > \lambda_1$  so that  $E_{\Lambda_1}$  exists.  $E_{\Lambda_1}$  is locally asymptotically stable for (13) if and only if  $f_2(N_1) < \frac{Z_1}{\eta_2} g'_2(0) + m_2$ ,  $\hat{A} > 0$  and  $\hat{A}\hat{B} > \hat{C}$ , where  $\hat{A}$ ,  $\hat{B}$ , and  $\hat{C}$  are as defined by equations (23).
4. Assume that  $N_0 - \frac{\Lambda_2 m_2}{\gamma_2 D} > N_2 > \lambda_2$  so that  $E_{\Lambda_2}$  exists.  $E_{\Lambda_2}$  is locally asymptotically stable for (13) if and only if  $f_1(N_2) < \frac{Z_2}{\eta_1} g'_1(0) + m_1$ ,  $\tilde{A} > 0$  and  $\tilde{A}\tilde{B} > \tilde{C}$ , where  $\tilde{A}$ ,  $\tilde{B}$ , and  $\tilde{C}$  are as defined by equations (22).
5. Suppose that  $N_0 = N_0^c$ , so that  $\bar{E} = (\lambda_2, \bar{P}_1, \bar{P}_2, 0)$  exists. Then  $\bar{E}$  is nonhyperbolic.
6. Suppose  $N_0 > N_0^c$ , so that there are two distinct equilibria of the form,  $\bar{E}' = (\lambda_2, \bar{P}'_1, \bar{P}'_2, 0)$  and  $\bar{E}'' = (\lambda_2, \bar{P}''_1, \bar{P}''_2, 0)$  with  $0 < \bar{P}'_1 < \bar{P}_1$ ,  $\bar{P}_2 < \bar{P}'_2 < 3\bar{P}_2/2$ ,  $\bar{P}_1 < \bar{P}''_1 < 3\bar{P}_1$ , and  $0 < \bar{P}''_2 < \bar{P}_2$ . Then  $\bar{E}''$  is a saddle, while  $\bar{E}' = (\lambda_2, \bar{P}'_1, \bar{P}'_2)$  satisfying (8), is locally asymptotically stable for (13) if and only if  $C > 0$ ,  $AB > C$ , and  $g_1(\bar{P}'_1) + g_2(\bar{P}'_2) < c$ . Here  $A$ ,  $B$ , and  $C$  are the same as in Theorem 2.1 (5).

**3.3. Global results.** In this subsection we list conditions under which  $E_0$ ,  $E_{\lambda_i}$ , and  $E_{\Lambda_i}$  are globally asymptotically stable for system (13) with respect to all solutions initiating in the positive cone of  $\mathbf{R}^4$ .

**Theorem 3.2.** If  $N_0 < \lambda_i$  for  $i = 1, 2$ , then the species-free steady state  $E_0$  is globally asymptotically stable for (13).

*Proof.* Since  $N_0 < \lambda_i$  for  $i = 1, 2$ , by Lemma 2.3 we obtain  $P_i(t) \rightarrow 0$  as  $t \rightarrow \infty$  in (13). Take

$$Q \in \{(N, P_1, P_2, Z) \in \mathbf{R}^4_+ : N > 0, P_1 > 0, P_2 > 0, Z > 0\}.$$

Let  $\Omega(Q)$  denote the omega limit set of the orbit through  $Q$ . Then it follows that any  $P = (N, P_1, P_2, Z) \in \Omega(Q)$  satisfies  $P_1 = 0, P_2 = 0$ . On  $\{(N, 0, 0, Z) \in \mathbf{R}^4_+\}$  the system reduces to

$$\begin{aligned} N'(t) &= (N_0 - N(t))D, \\ Z'(t) &= -cZ \end{aligned} \tag{24}$$

and hence  $N(t) \rightarrow N_0$  and  $Z(t) \rightarrow 0$  as  $t \rightarrow \infty$ . Therefore,  $\{E_0\} \in \Omega(Q)$ . Since all solutions of (13) are positive and bounded,  $\Omega(Q)$  is a nonempty compact subset of  $\mathbf{R}^4_+$ . If  $P \in \Omega(Q)$  then the entire trajectory through  $P$  is in  $\Omega(Q)$ . Hence,  $E_0$  is the only candidate. Thus,  $E_0$  is globally asymptotically stable for (13).  $\square$



The next theorem gives conditions under which  $E_{\lambda_i}$  is globally asymptotically stable for (13).

**Theorem 3.3.** (a) Suppose  $\lambda_1 < N_0 < \lambda_2$  and  $\omega_1 < c$ . Then  $E_{\lambda_1}$  is globally asymptotically stable for (13).

(b) Suppose  $\lambda_2 < N_0 < \lambda_1$  and  $\omega_2 < c$ . Then  $E_{\lambda_2}$  is globally asymptotically stable for (13).

*Proof.* We prove only (b) as the proof of (a) is symmetrical. Take

$$Q \in \{(N, P_1, P_2, Z) \in \mathbf{R}_+^4 : N > 0, P_1 > 0, P_2 > 0, Z > 0\}.$$

Let  $\Omega(Q)$  denote the omega limit set of the orbit through  $Q$ . Then it follows from Lemma 2.3 that any  $P = (N, P_1, P_2, Z) \in \Omega(Q)$  satisfies  $P_1 = 0$ . On  $\{(N, 0, P_2, Z) \in \mathbf{R}_+^4\}$  the system reduces to

$$\begin{aligned} N'(t) &= (N_0 - N(t))D - \frac{P_2(t)}{\gamma_2} f_2(N(t)), \\ P_2'(t) &= P_2(t)(f_2(N(t)) - m_2) - \frac{g_2(P_2)Z}{\eta_2}, \\ Z'(t) &= (g_2(P_2(t)) - c)Z(t). \end{aligned} \tag{25}$$

The third equation of (25) gives us

$$Z'(t) = (g_2(P_2(t)) - c)Z(t) < (\omega_2 - c)Z(t) \tag{26}$$

for all sufficiently large  $t$ . Hence,  $Z(t) \leq Z(0) \exp((\omega_2 - c)t) \rightarrow 0$  as  $t \rightarrow \infty$  (because  $\omega_2 < c$ ). Since  $Z(t) > 0$  we get  $Z(t) \rightarrow 0$  in (13). On  $\{(N, 0, P_2, 0) \in \mathbf{R}_+^4\}$  system (25) reduces to

$$\begin{aligned} N'(t) &= (N_0 - N(t))D - \frac{P_2(t)}{\gamma_2} f_2(N(t)), \\ P_2'(t) &= P_2(t)(f_2(N(t)) - m_2). \end{aligned} \tag{27}$$

By an argument comparable to that given in Hsu [21],  $N(t) \rightarrow \lambda_1$  and  $P_2(t) \rightarrow \bar{P}_2 = \frac{\gamma_2(N_0 - \lambda_2)D}{m_2}$ . Therefore,  $\{E_{\lambda_2}\} \in \Omega(Q)$ . Since (27) has no periodic orbits and the boundary is acyclic, it follows from Lemma 4.3 in Thieme [45] that  $E_{\lambda_2}$  is globally asymptotically stable for (13).  $\square$

The last result of this section gives conditions under which  $E_{\Lambda_i}$  is globally asymptotically stable for (13).

**Theorem 3.4.** (a) Assume that  $N_0 - \frac{\Lambda_1 m_1}{\gamma_1 D} > N_1 > \lambda_1$ ,  $f_2(N_1) < \frac{Z_1}{\eta_2} g_2'(0) + m_2$ , and  $\omega_1 > c$ . Then  $E_{\Lambda_1}$  is globally asymptotically stable for (13).

(b) Assume that  $N_0 - \frac{\Lambda_2 m_2}{\gamma_2 D} > N_2 > \lambda_2$ ,  $f_1(N_2) < \frac{Z_2}{\eta_1} g_1'(0) + m_1$ , and  $\omega_2 > c$ . Then  $E_{\Lambda_2}$  is globally asymptotically stable for (13).

*Proof.* We prove only (a) as the proof of (b) is symmetrical. Take

$$Q \in \{(N, P_1, P_2, Z) \in \mathbf{R}_+^4 : N > 0, P_1 > 0, P_2 > 0, Z > 0\}.$$

Let  $\Omega(Q)$  denote the omega limit set of the orbit through  $Q$ . Then it follows from Lemma 2.3 that any  $P = (N, P_1, P_2, Z) \in \Omega(Q)$  satisfies  $P_2 = 0$ . On

$\{(N, P_1, 0, Z) \in \mathbf{R}_+^4\}$  the system reduces to

$$\begin{aligned} N'(t) &= (N_0 - N(t))D - \frac{P_1(t)}{\gamma_1} f_1(N(t)), \\ P_1'(t) &= (f_1(N(t)) - m_1)P_1(t) - \frac{g_1(P_1)Z}{\eta_1}, \\ Z'(t) &= (g_1(P_1(t)) - c)Z(t). \end{aligned} \tag{28}$$

Applying Lemma 2.2 to the third equation of (28) gives us

$$(g_2(P_2(t)) - c)Z(t) \rightarrow 0, \tag{29}$$

as  $t \rightarrow \infty$ . Given that  $\omega_1 > c$ ,  $Z(t)$  cannot approach 0 as  $t \rightarrow \infty$ . Hence,  $P_1(t) \rightarrow \Lambda_1$  in (13). On  $\{(N, \Lambda_1, 0, Z) \in \mathbf{R}_+^4\}$  system (28) reduces to

$$\begin{aligned} N'(t) &= (N_0 - N(t))D - \frac{\Lambda_1}{\gamma_1} f_1(N(t)), \\ P_1'(t) &= (f_1(N(t)) - m_1)\Lambda_1 - \frac{\Lambda_1 h_1(\Lambda_1)Z}{\eta_1}, \end{aligned} \tag{30}$$

because  $g_1(P_1) = P_1 h_1(P_1)$  by assumptions on the response function  $g_1$ . By an argument comparable to that given in Hsu [21],  $N(t) \rightarrow N_1$  and  $Z(t) \rightarrow Z_1 = \frac{\eta_1(f_1(N_1) - m_1)}{h_1(\Lambda_1)}$ . Therefore,  $\{E_{\Lambda_1}\} \in \Omega(Q)$ . Since (30) has no periodic orbits and the boundary is acyclic, it follows from Lemma 4.3 in Thieme [45] that  $E_{\Lambda_1}$  is globally asymptotically stable for (13).  $\square$

We found two new equilibria of the form  $(N_1, \Lambda_1, 0, Z_1)$  and  $(N_2, 0, \Lambda_2, Z_2)$  that did not exist for the model (1). These new equilibria are globally asymptotically stable under conditions given by Theorem 3.4. Another striking change on model (1) under predation (referred to as model (13)) is the existence and uniqueness of an interior equilibrium point (see Figure 7). The analytical part of this result is achieved through an appropriate choice of the model parameters and variables. The proof is left for future investigation. This shows that adding predation to model (1) increases the number of boundary equilibria and brings diversity in the ecosystem.

Note that it is a simple matter to incorporate nutrient recycling into system (13) in much the same way this was done for model (1). Although delays in nutrient recycling may occur in some systems (see for example, Beretta *et al.* [5]), we assume instantaneous recycling of nutrients lost from phytoplankton and zooplankton. Grover and Holt [13] justify this assumption by the statement that nutrient recycling results from microorganisms with shorter generation times and more rapid metabolism than other biotic components. Model (13) under instantaneous nutrient

recycling takes the form:

$$\begin{aligned}
 \frac{dN}{dt} &= (N_0 - N)D - \frac{1}{\gamma_1}f_1(N)P_1 - \frac{1}{\gamma_2}f_2(N)P_2 \\
 &\quad + \theta_1(\epsilon_1 + \gamma P_1 P_2^2)P_1 + \theta_2(\epsilon_2 + a_2)P_2 + \theta_3 \xi Z, \\
 \frac{dP_1}{dt} &= f_1(N)P_1 - m_1 P_1 - \frac{1}{\eta_1}g_1(P_1)Z - \gamma P_1^2 P_2^2, \\
 \frac{dP_2}{dt} &= f_2(N)P_2 - m_2 P_2 - \frac{1}{\eta_2}g_2(P_2)Z, \\
 \frac{dZ}{dt} &= (g_1(P_1) + g_2(P_2) - c)Z, \\
 N(0) &> 0, P_1(0) > 0, P_2(0) > 0, Z(0) > 0.
 \end{aligned}
 \tag{31}$$

In these equations  $\theta_1$ ,  $\theta_2$ , and  $\theta_3$  (assumed to be less than 1) are the nutrient contents of a single cell of species  $P_1$ ,  $P_2$ , and  $Z$ , respectively. The variables and remaining parameters of model (31) are the same as in model (13). It is straightforward to verify that instantaneous nutrient recycling does not impact the qualitative behavior of model (13). Only the values at which transitions take place are altered. The explanations are similar to those for model (11) and therefore are omitted.

#### 4. Examples.

4.1. **Example 1.** In the figures of this subsection, each of the outcomes previously described analytically is illustrated. We ran simulations using Matlab. We consider the system (1) with  $D = m_i = \gamma_i = 1$ ,  $i = 1, 2$ ,  $\gamma = 0.11$ ,  $\mu_1 = 8.5$ ,  $\mu_2 = 6$ ,  $K_1 = 0.6$ , and  $K_2 = 0.7$ . The consumption functions  $f_i$  follow Michaelis-Menten kinetics, so that  $f_1(N) = \frac{8.5N}{0.6+N}$  and  $f_2(N) = \frac{6N}{0.7+N}$ . It is straightforward to check that  $\lambda_1 = 0.07$  and  $\lambda_2 = 0.14$ . The time frame for our simulations was 0 to 25,000 days. Only the first 60 days are shown in the figures of this subsection. A very large grid of initial conditions  $(N(0), P_1(0), P_2(0))$  in the positive cone  $Int(\mathbf{R}_+^3)$  was used to ensure convergence of all solutions of system (1) to the indicated boundary equilibrium. For illustration, we use  $(N(0), P_1(0), P_2(0)) = (0.1, 0.7, 0.7)$ .

We first take  $N_0 = 0.05$ , so that  $N_0 < \lambda_1 < \lambda_2$ . By Theorem 2.2, the species-free steady state  $E_0$  is globally asymptotically stable for (1): all solutions of (1) tend to  $E_0$  regardless of initial condition. One such solution is depicted in Figure 2.

We then increase  $N_0$  to 0.1, so that  $\lambda_1 < N_0 < \lambda_2$ . By Theorem 2.3,  $E_{\lambda_1}$  is globally asymptotically stable for (1): all solutions of (1) tend to  $E_{\lambda_1}$  regardless of initial condition. One such solution is depicted in Figure 3.

4.2. **Example 2.** In this subsection the outcomes for the competition-allelopathy-predation model are illustrated. We ran simulations using Mathematica Version 10. We consider the system (13) with  $D = m_i = \gamma_i = \eta_i = 1$ ,  $i = 1, 2$ ,  $c = 1.1$ ,  $\mu_1 = 8.5$ ,  $\mu_2 = 6$ ,  $K_1 = 0.6$ ,  $K_2 = 0.7$ ,  $\omega_1 = 8.4$  (except in Figure 5 where  $\omega_1 = 1$ ),  $\omega_2 = 5.9$ ,  $L_1 = 0.9$ , and  $L_2 = 1.8$ . The phytotoxic coefficient is taken to be  $\gamma = 0.11$ . The consumption functions  $f_i$  and  $g_i$  follow Michaelis-Menten kinetics, so that  $f_1(N) = \frac{8.5N}{0.6+N}$ ,  $f_2(N) = \frac{6N}{0.7+N}$ ,  $g_1(P_1) = \frac{8.4P_1}{0.9+P_1}$ , and  $g_2(N) = \frac{5.9P_2}{1.8+P_2}$ . The time frame for our simulations was 0 to 25,000 days. Only the first 60 or 400 days are shown in the figures of this subsection. In all figures we take  $(N(0), P_1(0), P_2(0), Z(0)) = (0.1, 0.7, 0.7, 0.8)$ . It is straightforward to check that  $\lambda_1 = 0.07$ ,  $\lambda_2 = 0.14$ ,  $\Lambda_1 = 0.1356$ ,  $\Lambda_2 = 0.7125$ ,  $N_1 = 0.2556$ , and  $N_2 = 0.0939$ .

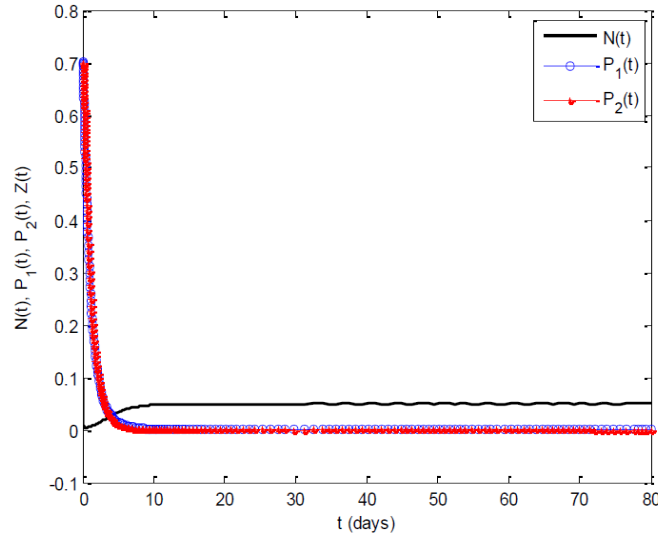


FIGURE 2. Graphs of  $N(t)$ ,  $P_1(t)$ , and  $P_2(t)$ . The input nutrient concentration  $N_0 = 0.05$ , satisfies  $N_0 < \lambda_1 < \lambda_2$ . By Theorem 2.2, the washout equilibrium  $E_0$  is globally asymptotically stable for model (1).

We first take  $N_0 = 0.05$ , so that  $N_0 < \lambda_1 < \lambda_2$ . By Theorem 3.2, the species-free steady state  $E_0$  is globally asymptotically stable for (13): all solutions of (13) tend to  $E_0$  regardless of initial condition. One such solution is depicted in Figure 4.

We then increase  $N_0$  to 0.12, so that  $\lambda_1 < N_0 < \lambda_2$ , and we pick  $\omega_1 = 1$ , so that  $\omega_1 < c$ . By Theorem 3.3,  $E_{\lambda_1}$  is globally asymptotically stable for (13): all solutions of (13) tend to  $E_{\lambda_1}$  regardless of initial data. One such solution is depicted in Figure 5.

We further increase  $N_0$  to 0.60, so that  $N_0 - \frac{\Lambda_1 m_1}{\gamma_1 D} > N_1 > \lambda_1$ , and  $f_2(N_1) < \frac{Z_1}{\eta_2} g_2'(0) + m_2$ . We choose  $\omega_1 = 8.4$ , so that  $\omega_1 > c$ . By Theorem 3.4,  $E_{\Lambda_1}$  is globally asymptotically stable for (13): all solutions of (13) tend to  $E_{\Lambda_1}$  regardless of initial conditions as shown in Figure 6.

We finally increase  $N_0$  to 0.70. Figure 7 shows that there is a unique interior equilibrium for model (13).

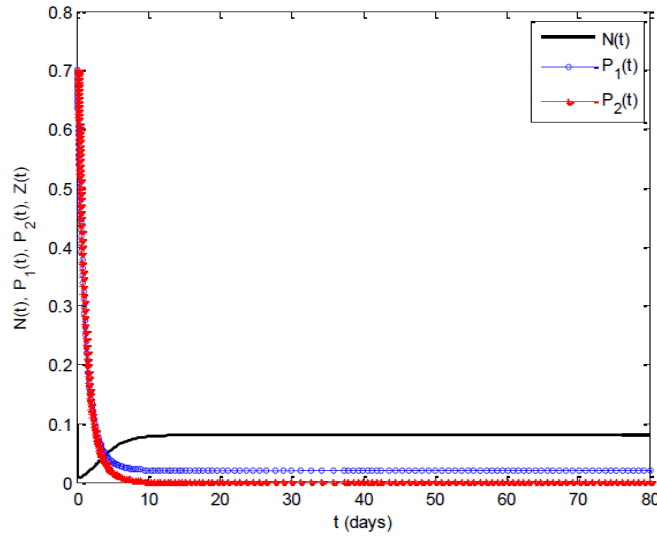


FIGURE 3. Graphs of  $N(t)$ ,  $P_1(t)$ , and  $P_2(t)$ . The input nutrient concentration  $N_0 = 0.10$  satisfies  $\lambda_1 < N_0 < \lambda_2$ . By Theorem 2.3 (a), the non-toxic monospecies  $E_{\lambda_1}$  is globally asymptotically stable for model (1).

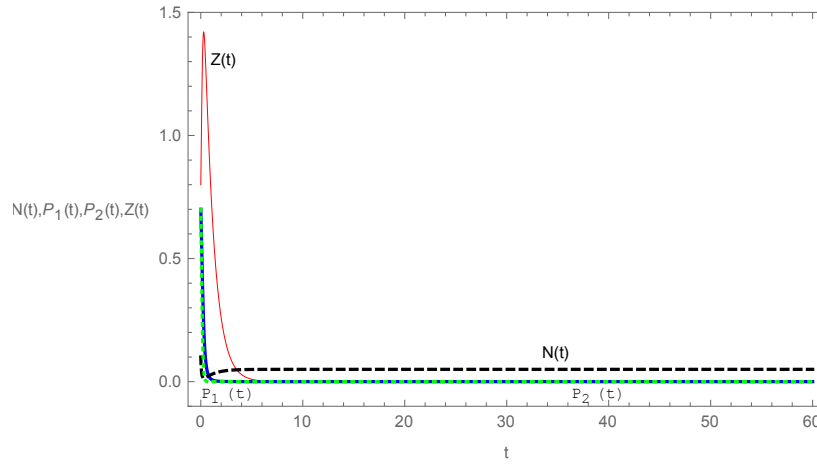


FIGURE 4. Graphs of  $N(t)$ ,  $P_1(t)$ ,  $P_2(t)$ , and  $Z(t)$ . The input nutrient concentration  $N_0 = 0.05$  satisfies  $N_0 < \lambda_1 < \lambda_2$ . By Theorem 3.2, the species-free equilibrium is globally asymptotically stable for model (13).

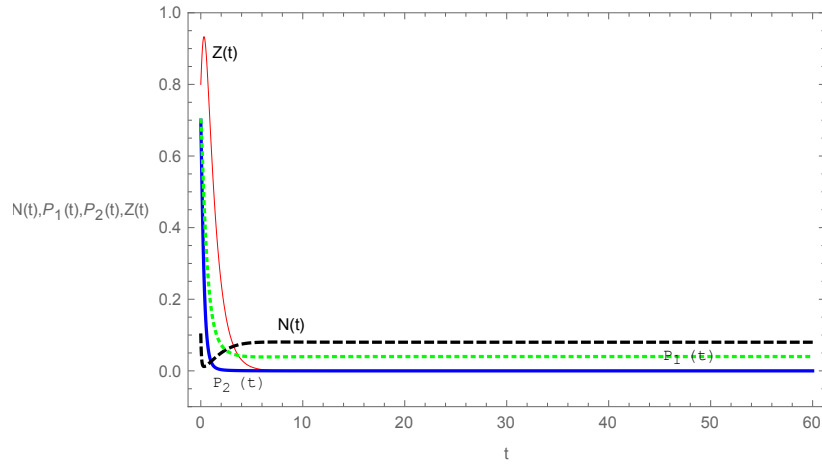


FIGURE 5. Graphs of  $N(t)$ ,  $P_1(t)$ ,  $P_2(t)$ , and  $Z(t)$ . The input nutrient concentration  $N_0 = 0.12$  satisfies  $\lambda_1 < N_0 < \lambda_2$ , and  $c > \omega_1$ , ( $c = 1.1$ ,  $\omega_1 = 1$ ). By Theorem 3.3 (a), the nontoxic monospecies equilibrium  $E_{\lambda_1}$  is globally asymptotically stable for model (13).

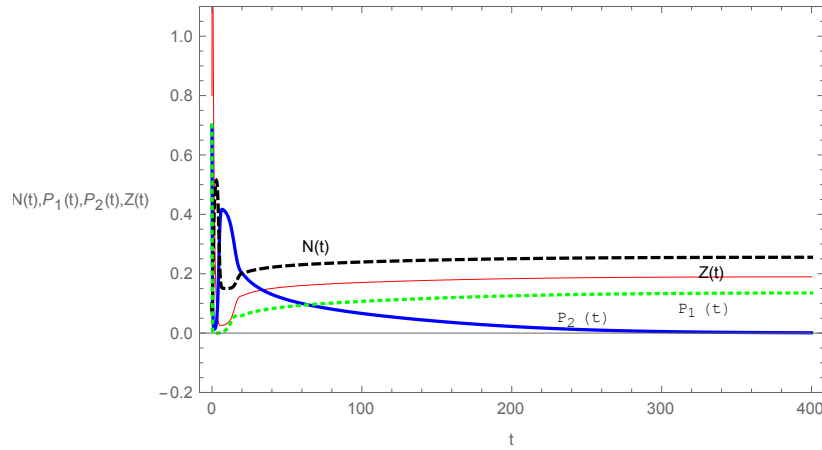


FIGURE 6. Graphs of  $N(t)$ ,  $P_1(t)$ ,  $P_2(t)$ , and  $Z(t)$ . The input nutrient concentration  $N_0 = 0.60$  satisfies  $N_0 - \frac{\Lambda_1 m_1}{\gamma_1 D} > N_1 > \lambda_1$ . Also,  $\omega_1 = 8.4 > c$  and  $f_2(N_1) < \frac{Z_1}{\eta_2} g_2'(0) + m_2$ . By Theorem 3.4 (a), the equilibrium  $E_{\Lambda_1}$  is globally asymptotically stable for model (13).

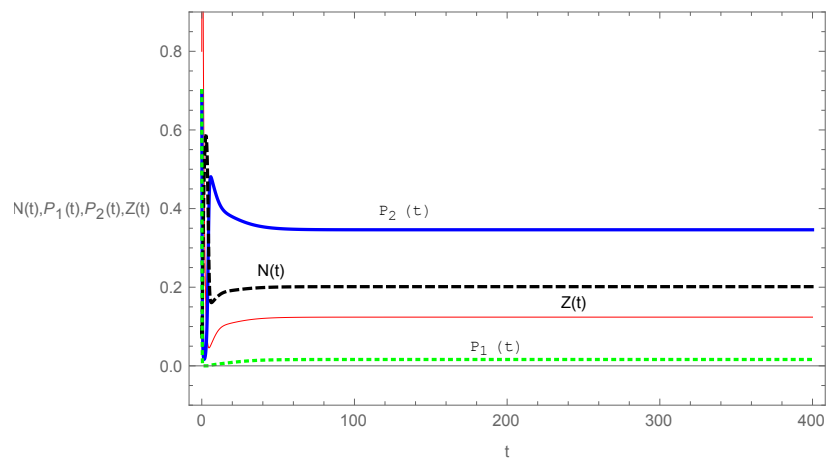


FIGURE 7. Graphs of  $N(t)$ ,  $P_1(t)$ ,  $P_2(t)$ , and  $Z(t)$ . The input nutrient concentration  $N_0 = 0.70$  satisfies the assumptions on the existence of an interior equilibrium of model (13).

**5. Concluding remarks.** In this paper, we have extended the work of Roy [38] to general uptake functions, demonstrating again that allelopathy due to one toxin-producing species can lead to stable coexistence (while competitive exclusion of the weaker competitor would otherwise occur). Roy used the phytotoxic coefficient  $\gamma$  for his bifurcation analysis. We have used the input nutrient concentration  $N_0$  as our bifurcation parameter because it can be controlled by the experimenter. As  $N_0$  is increased, the model exhibits a biologically relevant evolution of equilibria into the nonnegative cone of  $\mathbf{R}^3$  and the consequent transfer of stability. We have studied local and global stability of each of the boundary equilibria  $E_0$ ,  $E_{\lambda_1}$ , and  $E_{\lambda_2}$  of model (1). When  $N_0$  reaches at least the threshold  $N_0^c$ , we are able to study local stability of interior equilibria. Model (1) cannot exhibit uniform persistence because when  $N_0 > N_0^c$ , we have two distinct interior equilibria: one is a saddle and the other one is locally asymptotically stable. This result remains true when the toxic phytoplankton  $P_2$  is the stronger competitor. We recall that when  $\lambda_2 < \lambda_1$  ( $P_2$  is stronger in competition than  $P_1$ ) there is no interior equilibrium.

The incorporation of nutrient recycling into the model (1) modifies only the parameter bounds at which transitions occur, leaving the qualitative dynamics of the model (1) unchanged.

Predation can be responsible for diversity in ecosystems. Predation may promote, hinder or have no effect on interspecific competitive interactions (Chesson *et al.* [10]). We introduced predation of phytoplankton species by zooplankton in the model (1) and analyzed the modified system. Model (13) illustrates how predation increases the number of boundary equilibria of model (1) and brings diversity to interspecific competitive interactions. When the input nutrient concentration  $N_0$  reaches a critical value, model (1) has one or two interior equilibria while our simulations indicate that model (13) has a unique interior equilibrium (see Figure 7). The system (31) extends model (13) to incorporate instantaneous nutrient recycling.

In a recent work (Kengwoung-Keumo [24]), we showed that model (13), in the absence of phytotoxic effects and nutrient recycling, exhibits uniform persistence. For future work, we could investigate if the persistence result still holds in the presence of phytotoxic effects and nutrient recycling. We could also incorporate excretion and/or the phytotoxic effects of toxins on the growth of zooplankton and study the impact(s) on the model (13).

**Acknowledgments.** The author sincerely appreciates the constructive comments and suggestions of the editor and two referees.

## REFERENCES

- [1] M. An, D. Liu, I. Johnson and J. Lovett, Mathematical modelling of allelopathy. II. The dynamics of allelochemicals from living plants in the environment, *Ecol. Model.*, **161** (2003), 53–66.
- [2] H. M. Anderson, V. Hutson and R. Law, On the conditions for persistence of species in ecological communities, *Amer. Natur.*, **139** (1992), 663–668.
- [3] R. Aris and A. E. Humphrey, Dynamics of a chemostat in which two organisms compete for a common substrate, *Biotechnol. Bioeng.*, **19** (1977), 707–723.
- [4] M. M. Ballyk and G. S. K. Wolkowicz, Exploitative competition in the chemostat for two perfectly substitutable resources, *Math. Biosci.*, **118** (1993), 127–180.
- [5] E. Beretta, Bischi and G. F. Solimano, Stability in chemostat equations with delayed nutrient recycling, *J. Math. Biol.*, **28** (1990), 99–111.
- [6] B. Boon and H. Laudelout, Kinetics of nitrite oxidation by *Nitrobacter winogradskyi*, *Biochem. J.*, **85** (1962), 440–447.



- [7] G. J. Butler and G. S. K. Wolkowicz, Predator-mediated competition in the chemostat, *J. Math. Biol.*, **24** (1986), 167–191.
- [8] G. J. Butler and G. S. K. Wolkowicz, Exploitative competition in a chemostat for two complementary, and possibly inhibitory, resources, *Math. Biosci.*, **83** (1987), 1–48.
- [9] S. Chakraborty and J. Chattopadhyay, Nutrient-phytoplankton-zooplankton dynamics in the presence of additional food source- A mathematical study, *J. Biol. Syst.*, **16** (2008), 547–564.
- [10] P. Chesson, J. M. Chase, P. A. Abrams, J. P. Grover, S. Diehl, R. D. Holt, S. A. Richards, R. M. Nisbet and T. J. Case, The interaction between predation and competition: A review and synthesis, *Eco. Let.*, **5** (2002), 302–315.
- [11] A. M. Edwards and J. Brindley, Zooplankton mortality and the dynamical behaviour of plankton population models, *Bull. Math. Biol.*, **6** (1999), 303–339.
- [12] A. M. Edwards, Adding detritus to a nutrient-phytoplankton-zooplankton model: A dynamical-systems approach, *J. Plankton Res.*, **23** (2001), 389–413.
- [13] J. P. Grover and R. D. Holt, Disentangling resource and apparent competition: Realistic models for plant-herbivore communities, *J. Theor. Biol.*, **191** (1998), 353–376.
- [14] T. G. Hallam, On persistence of aquatic ecosystems, In: *Anderson, N. R., Zahuranc, B. G. (eds). Oceanic Sound Scattering Predication*, 1977, 749–765. New York: Plenum.
- [15] T. G. Hallam, Controlled persistence in rudimentary plankton models, In: *Avula, J. R. (eds). Math. Model.*, **4** (1977), 2081–2088. Rolla: University of Missouri Press.
- [16] T. G. Hallam, Structural Sensitivity of grazing formulation in nutrient controlled plankton models, *J. Math. Biol.*, **5** (1978), 261–280.
- [17] S. R. Hansen and S. P. Hubbell, Single-nutrient microbial competition: Qualitative agreement between experimental and theoretical forecast outcomes, *Sci.*, **207** (1980), 1491–1493.
- [18] G. Hardin, The competitive exclusion principle, *Sci.*, New series, **131** (1960), 1292–1297.
- [19] R. D. Holt, J. Grover and D. Tilman, Simple rules for interspecific dominance in systems with exploitative and apparent competition, *Amer. Natur.*, **144** (1994), 741–771.
- [20] S. B. Hsu, S. Hubbell and P. Waltman, A mathematical theory for single-nutrient competition in continuous cultures of micro-organisms, *SIAM J. Appl. Math.*, **32** (1977), 366–383.
- [21] S. B. Hsu, Limiting behavior for competing species, *SIAM J. Appl. Math.*, **34** (1978), 760–763.
- [22] S. R. J. Jang and J. Baglama, Nutrient-plankton models with nutrient recycling, *Comput. Math. Appl.*, **49** (2005), 375–378.
- [23] J. L. Jost, S. F. Drake, A. G. Fredrickson and M. Tsuchiya, Interaction of tetrahymena pyriformis, escherichia, coli, azotobacter vinelandii and glucose in a minimal medium, *J. Bacteriol.*, **113** (1976), 834–840.
- [24] J.-J. Kengwoung-Keumo, Dynamics of two phytoplankton populations under predation, *J. Math. Bio. Eng.*, **11** (2014), 1319–1336.
- [25] J. A. León and D. B. Tumpson, Competition between two species for two complementary or substitutable resources, *J. Theor. Biol.*, **50** (1975), 185–201.
- [26] B. Li and Y. Kuang, Simple Food Chain in a Chemostat with Distinct Removal Rates, *J. Math. Anal. Appl.*, **242** (2000), 75–92.
- [27] J. Maynard-Smith, *Models in Ecology*, Cambridge University Press, 1974.
- [28] R. K. Miller, *Nonlinear Volterra Equation*, W. A. Benjamin, N.Y., 1971.
- [29] H. Molisch, *Der Einfluss Einer Pflanze Auf Die Andere-Allelopathie*, Fischer, Jena, 1937.
- [30] J. Monod, *Recherche Sur La Croissance Des Cultures Bacteriennes*, Hermann et Cie., Paris, 1942.

- [31] B. Mukhopadhyay and R. Bhattacharyya, Modelling phytoplankton allelopathy in a nutrient-plankton model with spatial heterogeneity, *Ecol. Model.*, **198** (2006), 163–173.
- [32] M. Nakamaru and Y. Iwasa, Competition by allelopathy proceeds in travelling waves: colicin-immune strain aids colicin-sensitive strain, *Theor. Popul. Biol.*, **57** (2000), 131–144.
- [33] L. Perko, *Differential Equations and Dynamical Systems*, Third Edition, Springer, 2001.
- [34] M. J. Piotrowska, U. Forys and M. Bodnar, A simple model of carcinogenic mutations with time delay and diffusion, *Math. Biosci. Eng.*, **10** (2013), 861–872.
- [35] D. Rapport, An optimization model of food selection, *Amer. Natur.*, **105** (1971), 575–587.
- [36] E. L. Rice, *Allelopathy*, Academic Press, Inc. 1984.
- [37] G. A. Riley, A mathematical model of regional variations in plankton, *Limnol. Oceanog.*, **10** (Suppl.) (1965), R202–R215.
- [38] S. Roy, The coevolution of two phytoplankton species on a single resource: Allelopathy as a pseudo-mixotrophy, *Theor. Popul. Biol.*, **75** (2009), 68–75.
- [39] P. K Roy, A. N. Chatterjee, D. Greenhalgh and D. Khan, Long term dynamics in a mathematical model of HIV-1 infection with delay in different variants of the basic drug therapy model, *Nonlinear Anal-Real.*, **14** (2013), 1621–1633.
- [40] S. Ruan, Persistence and coexistence in zooplankton-phytoplankton-nutrient models with instantaneous nutrient recycling, *J. Math. Biol.*, **31** (1993), 633–654.
- [41] S. Ruan, Oscillations in plankton models with recycling, *J. Theor. Biol.*, **208** (2001), 15–26.
- [42] A. Sinkkonen, Modelling the effect of autotoxicity on density-dependent phytotoxicity, *J. Theor. Biol.*, **244** (2007), 218–227.
- [43] H. L. Smith and P. Waltman, *The theory of the chemostat: Dynamics of microbial competition*, Vol. 13 of Cambridge Studies in Mathematical Biology, Cambridge University Press, Cambridge, U.K., 1995.
- [44] J. Solé, E. Garcìa-Ladona, P. Ruardij and M. Estrada, Modelling allelopathy among marine algae, *Ecol. Model.*, **183** (2005), 373–384.
- [45] H. R. Thieme, Convergence results and a Poincaré-Bendixson trichotomy for asymptotically autonomous differential equations, *J. Math. Biol.*, **30** (1992), 755–763.
- [46] H. V. Thurman, *Introductory Oceanography*, 8th edition, Englewood Cliffs, NJ: Prentice-Hall, 1997.
- [47] R. Whittaker and P. Feeny, Allelochemicals: chemical interactions between species, *Sci.*, **171** (1971), 757–770.
- [48] G. S. K. Wolkowicz and Z. Lu, Global dynamics of a mathematical model of competition in the chemostat: General response functions and differential death rates, *J. Appl. Math.*, **52** (1992), 222–233.
- [49] R. D. Yang and A. E. Humphrey, Dynamics and steady state studies of phenol biodegradation in pure and mixed cultures, *Biotechnol. Bioeng.*, **17** (1975), 1211–1235.

Received May 13, 2015; Accepted December 30, 2015.

E-mail address: [jkengwou@cameron.edu](mailto:jkengwou@cameron.edu)