### Killer Yeast Vs. Sensitive Yeast

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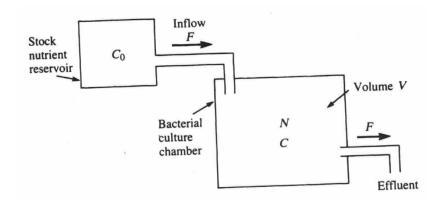
MATH 445 - Statistical, Dynamical, and Computational Modeling

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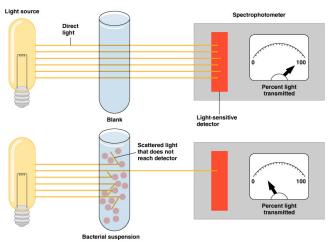
## Experiment

An experiment was conducted where an amount of yeast was grown in *chemostat* and allowed to come to equilibrium. The steady-state *optical density* of the yeast was recorded for a given flow rate F.

### Chemostat



## Optical Density



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### ODE

The differential equation we may use for modeling the growth of yeast is the same as that used for bacterial growth in a chemostat:

$$\begin{split} \frac{dN}{dt} &= k(C)N - \frac{FN}{V}, \\ \frac{dC}{dt} &= -\alpha k(C)N - \frac{FC}{V} + \frac{FC_0}{V}, \end{split}$$

with initial conditions  $C(0) = C_i$  and  $N(0) = N_i$ , N is the unitless optical density of yeast in the chamber, C is the unitless optical density of nutrient in the chamber,  $C_0$  is the unitless optical density of nutrient in the reservoir, F is the in/out volume flow rate with units volume/time, V is the volume of the chamber,  $\alpha$  is a unitless inverse of the yield constant, and k(C) is the reproduction rate for yeast in units 1/time.

# k(C) and its Simplification

k(C) is the reproduction rate for yeast in units 1/time with possible formula chosen such that  $\lim_{C\to\infty} = k_{max}$ , and  $k_{max}$  represents the maximum possible reproduction rate:

$$k(C) = \frac{k_{max}C}{C_n + C}.$$

where  $C_n$  is chosen such that  $k(C_n) = k_{max}/2$ . Because the concentration in the tank C(t) is related to the concentration in the reservoir by  $C(t) \leq C_0$ ,  $C_0$  may be chosen sufficiently small such that

$$k(C) = \frac{k_{max}C}{C_n + C} \approx \frac{k_{max}C}{C_n} = KC,$$

with K in units 1/time.

### New ODE

The equations we need to solve then become

$$\frac{dN}{dt} = KCN - \frac{FN}{V},\tag{1}$$

$$\frac{dN}{dt} = KCN - \frac{FN}{V}, \qquad (1)$$

$$\frac{dC}{dt} = -\alpha KCN - \frac{FC}{V} + \frac{FC_0}{V}. \qquad (2)$$

### Steady States

The quantitative measurement for fitness is a unitless measurement of optical density at steady state (N) at a given flow rate (F) in volumes/hr. In order to find the steady states, we have to find the intersections of the null-clines at equilibrium points  $(\bar{N}, \bar{C})$ , i.e.  $dN(\bar{N}, \bar{C})/dt = 0$  and  $dC(\bar{N}, \bar{C})/dt = 0$ :

# $dN(\bar{N}, \bar{C})/dt = 0$

$$\begin{split} \frac{dN(\bar{N},\bar{C})}{dt} &= K\bar{C}\bar{N} - \frac{F\bar{N}}{V}, \\ &= \bar{N}\left(K\bar{C} - \frac{F}{V}\right) = 0, \end{split}$$

which is zero for  $\bar{N} = 0$  or  $K\bar{C} = F/V$ .

$$dC(\bar{N}, \bar{C})/dt = 0$$

$$\frac{dC(\bar{N},\bar{C})}{dt} = -\alpha K\bar{C}\bar{N} - \frac{F\bar{C}}{V} + \frac{FC_0}{V} = 0,$$

which is zero for  $\alpha K \bar{C} \bar{N} + \frac{F\bar{C}}{V} = \frac{FC_0}{V}$ .

# Determination of $(\bar{N}, \bar{C})$

In order to evaluate these null-clines, we need to evaluate the non-trivial cases, here for  $\dot{N}=0$ ,

$$K\bar{C} = \frac{F}{V} \implies \bar{C} = \frac{F}{VK}.$$
 (3)

Likewise, for  $\dot{C} = 0$ ,

$$\alpha K \bar{C} \bar{N} + \frac{F \bar{C}}{V} = \frac{F C_0}{V}$$

$$\implies \bar{N} = \frac{F C_0}{V \alpha K \bar{C}} - \frac{F}{V \alpha K} = \frac{F}{V \alpha K} \left(\frac{C_0}{\bar{C}} - 1\right). \tag{4}$$

This intersects the  $\bar{N}=0$  nullcline at  $\frac{F}{V\alpha K}=0$  or  $\frac{C_0}{\bar{C}}=1$ . However, because F is never 0, we can disregard the first equation, and we know that the only trivial steady-state is located at  $\bar{N}=0$ ,  $\bar{C}=C_0$ .

# Determination of $(\bar{N}, \bar{C})$

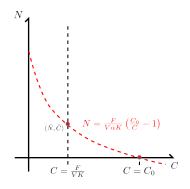


Figure: The  $\dot{C}=0$  nullcline (dashed red) intersecting with the  $\dot{N}=0$  nullcline (dashed black). The trivial and non-trivial steady-states  $(0,C_0)$  and  $(\bar{N},\bar{C})$  are shown as red dots.

By placing Eq. (3) inside Eq. (4), we can find the non-trivial steady-state, the intersection of null-clines:

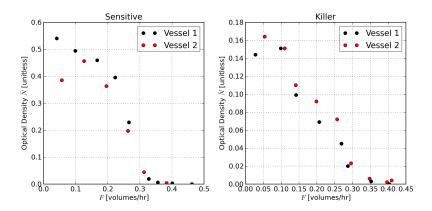
$$\bar{N}(\bar{C}) = \frac{FC_0}{V\alpha K\bar{C}} - \frac{F}{V\alpha K}$$

$$= \frac{FC_0}{V\alpha K\frac{F}{VK}} - \frac{F}{V\alpha K}$$

$$= \frac{C_0}{\alpha} - \frac{F}{V\alpha K} = \frac{1}{\alpha} \left( C_0 - \frac{F}{VK} \right). \quad (5)$$

The unknown parameters in Eq (5) are  $\alpha$  and K. We can find these parameters by fitting Eq (5) to the data by non-linear least squares fitting  $\bar{N}_i$  at  $F_i$  for  $i=1,\ldots,n$ , where n is the number of observations.

### The Data



$$C_0 = 0.02$$

# Eq. (5) Fit for Sensitive Yeast

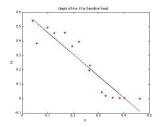


Figure: Sensitive yeast S steady-state best-fit line using Eq. (5).

	Estimates	SE	CI
$\alpha_S$	0.0325	0.0024	(0.0255, 0.0399)
$K_S$	20.1818	1.0802	(16.9281, 23.4356)

The results on the left show that there may be a better fit to the data than Eq. (5). Notice that elimination of the last few data points corresponding to high F may be removed; this would reduce the standard error and hence provide a better estimate for  $\alpha_L$  and  $K_L$ .

# Eq. (5) Fit for Killer Yeast

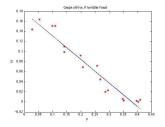


Figure: Killer yeast L steady-state best-fit line using Eq. (5).

	Estimates	SE	CI
$\alpha_L$	0.1124	0.0053	(0.0969, 0.1279)
$K_L$	19.0288	0.6367	(17.1526, 20.905)

The best-fit line on the left shows that steady-state data follows a fairly linear relationship with flow, and as such we can be confident our estimates for  $\alpha_L$  and  $K_L$  are correct.

### What-if Scenario

Now that we have obtained estimates of the parameters for both the killer yeast L and sensitive yeast S, ( $\alpha_L$ ,  $\alpha_S$  and  $K_L$ ,  $K_S$  respectively), we model a "what if" scenario whereby we place both species of yeast, sensitive and killer, into one chemostat. The differential equations we use to solve this three-species model is

$$\frac{dL}{dt} = K_L C L - \frac{FL}{V},\tag{6}$$

$$\frac{dS}{dt} = K_S C S - \frac{FS}{V} - \beta S L,\tag{7}$$

$$\frac{dC}{dt} = -\alpha_L K_L C L - \alpha_S K_S C S - \frac{FC}{V} + \frac{FC_0}{V}.$$
 (8)

### What-if Scenario

We can run the dynamic model (6), (7), and (8) to equilibrium. By letting  $\beta$  range from 0 to 0.5 (in units 1/time), and F range from 0 to 0.5 (in units volume/time), we can run the model for each  $\beta$  and F to determine the regions in the  $\beta$ , F plane where sensitive yeast S overtake the killer yeast L and vice versa.

### What-if Scenario Results

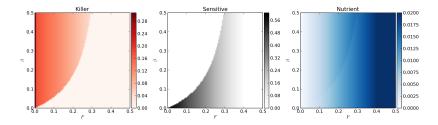


Figure:  $250 \times 250$  steady-state optical density solution for killer yeast (left), sensitive yeast (middle) and nutrient (right) for a total run time of 40,000 hours. Equations (6), (7), and (8) were solved with the Dormand-Prince numerical integration algorithm with an absolute tolerance of 1e-6, relative tolerance of 1e-6, and timestep  $\Delta t$  of 500 hours. The timestep was kept high due to the  $250 \times 250 = 62,500$  simulations required to complete the figure. Parallel processing was also implemented to speed up the simulation.

### What-if Scenario Results

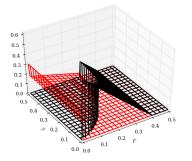


Figure: 3D solution depicting killer yeast (red) and sensitive yeast (black) steady-states.

Notice that as  $\beta$  increases the killer yeast dominates and sensitive yeast eventually dies off, while as F increases the sensitive yeast dominates and the killer yeast eventually dies out. The sensitive yeast are flushed from the container at around F=0.4 volumes/hr.

### Questions?

