# Nonlinear Systems of ODE: Dimensional Analysis, Equilibrium Points, and Stability 

MATH 469, Texas A\&M University

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## Non-dimensionalization

This lecture will focus on the competition model

$$
\begin{array}{ll}
\frac{d y_{1}}{d t}=r_{1} y_{1}\left(1-\frac{y_{1}+s_{1} y_{2}}{K_{1}}\right) ; & y_{1}(0)=y_{1_{0}} \\
\frac{d y_{2}}{d t}=r_{2} y_{2}\left(1-\frac{s_{2} y_{1}+y_{2}}{K_{2}}\right) ; & y_{2}(0)=y_{2_{0}},
\end{array}
$$

and we'll begin by non-dimensionalizing it. For this, we introduce three dimensionless variables

$$
\tau=\frac{t}{A} ; \quad Y_{1}(\tau)=\frac{y_{1}(t)}{B} ; \quad Y_{2}(\tau)=\frac{y_{2}(t)}{C}
$$

The constant $A$ will be chosen with the dimension time, and the constants $B$ and $C$ will both be chosen with dimension biomass.

## Non-dimensionalization

First, using the chain rule, we compute

$$
\begin{aligned}
& \frac{d y_{1}}{d t}=B \frac{d}{d t} Y_{1}(\tau)=B \frac{d}{d \tau} Y_{1}(\tau) \frac{d \tau}{d t}=\frac{B}{A} \frac{d Y_{1}}{d \tau} \\
& \frac{d y_{2}}{d t}=C \frac{d}{d t} Y_{2}(\tau)=C \frac{d}{d \tau} Y_{2}(\tau) \frac{d \tau}{d t}=\frac{C}{A} \frac{d Y_{2}}{d \tau}
\end{aligned}
$$

If we now substitute these dimensionless variables into the competition model, we get

$$
\begin{array}{ll}
\frac{B}{A} \frac{d Y_{1}}{d t}=r_{1} B Y_{1}\left(1-\frac{B Y_{1}+s_{1} C Y_{2}}{K_{1}}\right) ; & Y_{1}(0)=\frac{y_{10}}{B} \\
\frac{C}{A} \frac{d Y_{2}}{d t}=r_{2} C Y_{2}\left(1-\frac{s_{2} B Y_{1}+C Y_{2}}{K_{2}}\right) ; & Y_{2}(0)=\frac{y_{20}}{C} .
\end{array}
$$

## Non-dimensionalization

We multiply by $A$ and divide by $B$ to arrive at

$$
\begin{array}{ll}
\frac{d Y_{1}}{d t}=r_{1} A Y_{1}\left(1-\frac{B Y_{1}+s_{1} C Y_{2}}{K_{1}}\right) ; & Y_{1}(0)=\frac{y_{10}}{B} \\
\frac{d Y_{2}}{d t}=r_{2} A Y_{2}\left(1-\frac{s_{2} B Y_{1}+C Y_{2}}{K_{2}}\right) ; & Y_{2}(0)=\frac{y_{20}}{C}
\end{array}
$$

As always, our goal is to choose the constants $A, B$, and $C$ in a way that simplifies the system, while also ensuring that they have the correct dimensions. We'll take

$$
A=\frac{1}{r_{1}} ; \quad B=K_{1} ; \quad C=K_{2} .
$$

The system becomes

$$
\begin{aligned}
& \frac{d Y_{1}}{d t}=Y_{1}\left(1-Y_{1}-\frac{s_{1} K_{2}}{K_{1}} Y_{2}\right) ; \quad Y_{1}(0)=\frac{y_{1_{0}}}{K_{1}} \\
& \frac{d Y_{2}}{d t}=\frac{r_{2}}{r_{1}} Y_{2}\left(1-\frac{s_{2} K_{1}}{K_{2}} Y_{1}-Y_{2}\right) ; \quad Y_{2}(0)=\frac{y_{2_{0}}}{K_{2}} .
\end{aligned}
$$

## Non-dimensionalization

Recall that one of the things that we accomplish with non-dimensionalization is that we identify useful combinations of parameters. In this case, we set

$$
a=\frac{s_{1} K_{2}}{K_{1}} ; \quad b=\frac{r_{2}}{r_{1}} ; \quad c=\frac{s_{2} K_{1}}{K_{2}} .
$$

This allows us to write our system in the form we'll use for analysis,

$$
\begin{array}{ll}
\frac{d Y_{1}}{d t}=Y_{1}\left(1-Y_{1}-a Y_{2}\right) ; & Y_{1}(0)=\frac{y_{1_{0}}}{K_{1}} \\
\frac{d Y_{2}}{d t}=b Y_{2}\left(1-c Y_{1}-Y_{2}\right) ; & Y_{2}(0)=\frac{y_{2_{0}}}{K_{2}}
\end{array}
$$

## Equilibrium Points

For a first-order autonomous system of ODE

$$
\frac{d \vec{y}}{d t}=\vec{f}(\vec{y}) ; \quad \vec{y}(0)=\overrightarrow{y_{0}}
$$

we say that $\hat{y} \in \mathbb{R}^{n}$ is an equilibrium point if

$$
\vec{f}(\hat{y})=0 .
$$

As with single equations, if $\vec{y}_{0}=\hat{y}$, the constant function $\vec{y}(t)=\hat{y}$ for all $t \in \mathbb{R}$ is a solution to the system (because both sides of the equation are 0 ).

## Equilibrium Points

Example. Let's find all equilibrium points for the non-dimensionalized competition model,

$$
\begin{aligned}
& \frac{d y_{1}}{d t}=y_{1}\left(1-y_{1}-a y_{2}\right) \\
& \frac{d y_{2}}{d t}=b y_{2}\left(1-c y_{1}-y_{2}\right) .
\end{aligned}
$$

We need to find $\hat{y}=\binom{\hat{y}_{1}}{\hat{y}_{2}}$ so that

$$
\begin{aligned}
& 0=\hat{y}_{1}\left(1-\hat{y}_{1}-a \hat{y}_{2}\right) \\
& 0=b \hat{y}_{2}\left(1-c \hat{y}_{1}-\hat{y}_{2}\right) .
\end{aligned}
$$

For the first equation, we can either have $\hat{y}_{1}=0$ or $\hat{y}_{1}=1-a \hat{y}_{2}$. If we substitute $\hat{y}_{1}=0$ into the second equation, we get

$$
0=b \hat{y}_{2}\left(1-\hat{y}_{2}\right) \Longrightarrow \hat{y}_{2}=0,1
$$

This gives us our first two equilibrium points: $\binom{0}{0}$ and $\binom{0}{1}$.

## Equilibrium Points

Next, we substitute $\hat{y}_{1}=1-a \hat{y}_{2}$ into the second equation, giving

$$
0=b \hat{y}_{2}\left(1-c\left(1-a \hat{y}_{2}\right)-\hat{y}_{2}\right) .
$$

We see that $\hat{y}_{2}=0$ solves this equation, and for the second solution we have

$$
0=1-c+(a c-1) \hat{y}_{2} \Longrightarrow \hat{y}_{2}=\frac{1-c}{1-a c} .
$$

In order to find the values of $\hat{y}_{1}$ associated with these values of $\hat{y}_{2}$, we substitute back into $\hat{y}_{1}=1-a \hat{y}_{2}$. For $\hat{y}_{2}=0$, we see that $\hat{y}_{1}=1$, while for $\hat{y}_{2}=\frac{1-c}{1-a c}$, we compute

$$
\hat{y}_{1}=1-a \frac{1-c}{1-a c}=\frac{1-a c-a+a c}{1-a c}=\frac{1-a}{1-a c} .
$$

This gives us our next two equilibrium points: $\binom{1}{0}$ and $\binom{\frac{1-a}{1-a c}}{\frac{1-c}{1-a c}}$.

## Equilibrium Points

In total, we have four equilibrium points

$$
\binom{0}{0}, \quad\binom{0}{1}, \quad\binom{1}{0}, \quad\binom{\frac{1-a}{1-a c}}{\frac{1-c}{1-a c}} .
$$

Recalling that we scaled $y_{1}$ by dividing by $K_{1}$, and scaled $y_{2}$ by dividing by $K_{2}$, we see that we can interpret these equilibrium points in the original variables as described on the next slide.

## Equilibrium Points

$\binom{0}{0}$ : Both species die out;
$\binom{0}{k_{2}}$ : Species $y_{1}$ dies out and species $y_{2}$ goes to its carrying capacity;
$\binom{K_{1}}{0}$ : Species $y_{2}$ dies out and species $y_{1}$ goes to its carrying capacity;
$\binom{\frac{K_{1}-s_{1} K_{2}}{1-s_{1} \Sigma_{2}}}{\frac{K_{2}-K_{1} K_{1}}{1-s_{1} s_{2}}}$ If $K_{1}-s_{1} K_{2}, K_{2}-s_{2} K_{1}$, and $1-s_{1} s_{2}$ are all positive, the two species reach an equilibrium in which neither dies out. In principle, the same statement is true if these quantities are all negative, but we'll see below that that case isn't as interesting.

## Equilibrium Points

For the fourth equilibrium point, let's interpret what positivity of these quantities $K_{1}-s_{1} K_{2}, K_{2}-s_{2} K_{1}$, and $1-s_{1} s_{2}$ corresponds with. First, $s_{1} K_{2}$ denotes the amount resources available to species $y_{1}$ that species $y_{2}$ consumes at its carrying capacity. The requirement that $K_{1}-s_{1} K_{2}>0$ asserts that species $y_{2}$ cannot use up the entirety of these resources, even at its carrying capacity.

The condition $K_{2}-s_{2} K_{1}>0$ can be interpreted similarly, with the roles of the species reversed.

For the condition $1-s_{1} s_{2}>0$, recall our example in which $y_{1}$ corresponds with a population of rabbits and $y_{2}$ corresponds with a population of deer. Focusing on resources available to rabbits (and possibly available to the deer), suppose each individual in the deer population uses up twice the amount of these resources as an individual in the rabbit population. Then $s_{1}=2$.

## Equilibrium Points

Correspondingly, we might expect that if we focus on resources available to the deer (and possibly available to the rabbits), then each individual in the rabbit population will use up half as much of these resources as an individual in the deer population. I.e., we would have $s_{2}=\frac{1}{2}$.

In this case, we would have $s_{1} s_{2}=1$.
But suppose individuals in the deer population have access to resources unavailable to the rabbits. Then it might be the case that an individual rabbit uses less than half the resources of an individual deer. In that case, we would have $s_{2}<\frac{1}{2}$, and so $s_{1} s_{2}<1$.

Generally, the condition $1-s_{1} s_{2}>0$ indicates that at least one of the species has access to resources that are unavailable to the other species.

## Stability

Suppose $\hat{y} \in \mathbb{R}^{n}$ is an equilibrium point for the ODE system

$$
\frac{d \vec{y}}{d t}=\vec{f}(\vec{y}) ; \quad \vec{y}(0)=\overrightarrow{y_{0}} .
$$

I.e., $\vec{f}(\hat{y})=0$. As usual, we want to understand what will happen if we start with $\overrightarrow{y_{0}}$ near $\hat{y}$. To understand this, we write

$$
\begin{equation*}
\vec{y}(t)=\hat{y}+\vec{z}(t), \tag{}
\end{equation*}
$$

and we'll analyze the behavior of $\vec{z}(t)$ as $t$ increases. If we substitute $\left(^{*}\right)$ into our ODE system, and assume $\vec{f}$ is differentiable at $\hat{y}$, we get

$$
\frac{d \vec{z}}{d t}=\vec{f}(\hat{y}+\vec{z})=\vec{f}(\hat{y})+\vec{f}^{\prime}(\hat{y}) \vec{z}+\vec{\epsilon}(\vec{z} ; \hat{y})
$$

where

$$
\lim _{|\vec{z}| \rightarrow 0} \frac{|\vec{\epsilon}(\vec{z} ; \hat{y})|}{|\vec{z}|}=0
$$

## Stability

Using the relation $\vec{f}(\hat{y})=0$ and observing that $\vec{\epsilon}(\vec{z} ; \hat{y})$ is small relative to $\vec{f}^{\prime}(\hat{y}) \vec{z}$, we obtain the linearized equation

$$
\begin{equation*}
\frac{d \vec{z}}{d t}=\vec{f}^{\prime}(\hat{y}) \vec{z} ; \quad \vec{z}(0)=\vec{y}_{0}-\hat{y} . \tag{**}
\end{equation*}
$$

This is a first-order linear system of ODE with constant coefficients, and so we know how to solve it using the eigenvalues and eigenvectors of $\vec{f}^{\prime}(\hat{y})$.
Let $\left\{\lambda_{j}\right\}_{j=1}^{n}$ denote the eigenvalues of $\overrightarrow{f^{\prime}}(\hat{y})$, and assume these values are distinct. In this case, we can associate the eigenvalues with a linearly independent collection of eigenvectors $\left\{\vec{v}_{j}\right\}_{j=1}^{n}$.

## Stability

Using these eigenvalues and eigenvectors, we can express the general solution to $\left({ }^{* *}\right)$ as

$$
\vec{z}(t)=\sum_{j=1}^{n} c_{j} e^{\lambda_{j} t} \vec{v}_{j}
$$

for some constants $\left\{c_{j}\right\}_{j=1}^{n}$.
Even if the entries of $\vec{f}^{\prime}(\hat{y})$ are all real, it may have complex eigenvalues,

$$
\lambda_{j}=\operatorname{Re} \lambda_{j}+i \operatorname{Im} \lambda_{j}
$$

In this case, the complex modulus of $e^{\lambda_{j} t}$ is

$$
\begin{aligned}
\left|e^{\lambda_{j} t}\right| & =\left|e^{t \operatorname{Re} \lambda_{j}+i t \operatorname{Im} \lambda_{j}}\right|=\left|e^{t \operatorname{Re} \lambda_{j}} e^{i t \operatorname{Im} \lambda_{j}}\right| \\
& =\left|e^{t \operatorname{Re} \lambda_{j}}\right|\left|e^{i t \operatorname{Im} \lambda_{j}}\right|=e^{t \operatorname{Re} \lambda_{j}}
\end{aligned}
$$

## Stability

In obtaining the final equality on the previous page, we observed that $e^{t R e \lambda}$ is always a positive real number, and

$$
\left|e^{i t \operatorname{Im} \lambda_{j}}\right|^{2}=e^{i t \operatorname{Im} \lambda_{j}} e^{-i t \operatorname{Im} \lambda_{j}}=1 \Longrightarrow\left|e^{i t \operatorname{Im} \lambda_{j}}\right|=1
$$

We see that:

- If $\operatorname{Re} \lambda_{j}<0$ for all $j=1,2, \ldots, n$, then $\hat{y}$ will be asymptotically stable;
- If $\operatorname{Re} \lambda_{j}>0$ for at least one $j \in\{1,2, \ldots, n\}$, then $\hat{y}$ will be unstable.
- If $\operatorname{Re} \lambda_{j} \leq 0$ for all $j=1,2, \ldots, n$, and $\operatorname{Re} \lambda_{j}=0$ for at least one $j \in\{1,2, \ldots, n\}$, then this test for stability is inconclusive.

These conditions remain valid if the eigenvalues $\left\{\lambda_{j}\right\}_{j=1}^{n}$ are not distinct.

## Stability

Example. For the non-dimensionalized competition model

$$
\begin{aligned}
& \frac{d y_{1}}{d t}=y_{1}\left(1-y_{1}-a y_{2}\right) \\
& \frac{d y_{2}}{d t}=b y_{2}\left(1-c y_{1}-y_{2}\right)
\end{aligned}
$$

with $a, b, c>0$, analyze the stability of each of the four equilibrium points

$$
\binom{0}{0}, \quad\binom{0}{1}, \quad\binom{1}{0}, \quad\binom{\frac{1-a}{1-a c}}{\frac{1-c}{1-a c}} .
$$

We begin by computing the Jacobian matrix $\vec{f}^{\prime}(\vec{y})$. First:

$$
\vec{f}(\vec{y})=\binom{f_{1}\left(y_{1}, y_{2}\right)}{f_{2}\left(y_{1}, y_{2}\right)}=\binom{y_{1}-y_{1}^{2}-a y_{1} y_{2}}{b y_{2}-b c y_{1} y_{2}-b y_{2}^{2}} .
$$

## Stability

This allows us to compute

$$
\vec{f}^{\prime}(\vec{y})=\left(\begin{array}{cc}
1-2 y_{1}-a y_{2} & -a y_{1} \\
-b c y_{2} & b-b c y_{1}-2 b y_{2}
\end{array}\right) .
$$

We can now use this to evaluate each of the four equilibrium points.
For $\hat{y}=\binom{0}{0}$, we have

$$
\vec{f}^{\prime}(0,0)=\left(\begin{array}{ll}
1 & 0 \\
0 & b
\end{array}\right)
$$

The eigenvalues are 1 and $b$, which are both positive values, so this equilibrium point is unstable.

For $\hat{y}=\binom{0}{1}$, we have

$$
\overrightarrow{f^{\prime}}(0,1)=\left(\begin{array}{cc}
1-a & 0 \\
-b c & -b
\end{array}\right) .
$$

## Stability

The eigenvalues of this matrix are $1-a$ and $-b$. Since $-b<0$, the only condition that this imposes is

$$
1-a<0 \Longrightarrow a>1
$$

In terms of the original parameters, this asserts that the equilibrium point $\binom{0}{K_{2}}$ will be asymptotically stable provided

$$
a=\frac{s_{1} K_{2}}{K_{1}}>1 \Longrightarrow s_{1} K_{2}>K_{1}
$$

I.e., the second species will win if at its carrying capacity it uses more resources than are available to the first species.

Here, keep in mind that when talking about stability, we're always talking about solutions that start near the equilibrium point. Strictly speaking, this tells us that if $s_{1} K_{2}>K_{1}$, and the populations are initially near $\binom{0}{K_{2}}$, then the second species will win.

## Stability

Recall that the Jacobian matrix is

$$
\vec{f}^{\prime}(\vec{y})=\left(\begin{array}{cc}
1-2 y_{1}-a y_{2} & -a y_{1} \\
-b c y_{2} & b-b c y_{1}-2 b y_{2}
\end{array}\right) .
$$

For $\hat{y}=\binom{1}{0}$, we have

$$
\overrightarrow{f^{\prime}}(1,0)=\left(\begin{array}{cc}
-1 & -a \\
0 & b-b c
\end{array}\right)
$$

The eigenvalues are -1 and $b(1-c)$. Since -1 is negative, the only condition this imposes is

$$
b(1-c)<0 \Longrightarrow c>1
$$

## Stability

In terms of the original parameters, this asserts that the equilibrium point $\binom{K_{1}}{0}$ will be asymptotically stable provided

$$
c=\frac{s_{2} K_{1}}{K_{2}}>1 \Longrightarrow s_{2} K_{1}>K_{2}
$$

We can interpret this precisely as we did the previous equilibrium point, with the roles of the two species reversed.

## Stability

Recall again the Jacobian matrix

$$
\vec{f}^{\prime}(\vec{y})=\left(\begin{array}{cc}
1-2 y_{1}-a y_{2} & -a y_{1} \\
-b c y_{2} & b-b c y_{1}-2 b y_{2}
\end{array}\right) .
$$

For the last equilibrium point $\hat{y}=\binom{\frac{1-a}{1-a c}}{\frac{1-c}{1-a c}}$, it's helpful to recall that the equations we solved to find these values were

$$
\begin{aligned}
& 1-\hat{y}_{1}-a \hat{y}_{2}=0 \\
& 1-c \hat{y}_{1}-\hat{y}_{2}=0 .
\end{aligned}
$$

If we use these relations while evaluating $\vec{f}^{\prime}(\hat{y})$, we obtain
$\vec{f}^{\prime}(\hat{y})=\left(\begin{array}{cc}1-2 \hat{y}_{1}-a \hat{y}_{2} & -a \hat{y}_{1} \\ -b c \hat{y}_{2} & b-b c \hat{y}_{1}-2 b \hat{y}_{2}\end{array}\right)=\left(\begin{array}{cc}-\hat{y}_{1} & -a \hat{y}_{1} \\ -b c \hat{y}_{2} & -b \hat{y}_{2}\end{array}\right)$.

## Stability

In this case, we compute the eigenvalues by writing

$$
\begin{aligned}
& \operatorname{det}\left(\begin{array}{cc}
-\hat{y}_{1}-\lambda & -a \hat{y}_{1} \\
-b c \hat{y}_{2} & -b \hat{y}_{2}-\lambda
\end{array}\right)=\left(-\hat{y}_{1}-\lambda\right)\left(-b \hat{y}_{2}-\lambda\right)-a b c \hat{y}_{1} \hat{y}_{2} \\
& =\lambda^{2}+\left(\hat{y}_{1}+b \hat{y}_{2}\right) \lambda+b(1-a c) \hat{y}_{1} \hat{y}_{2}=0 .
\end{aligned}
$$

Solving this with the quadratic formula, we find

$$
\lambda_{ \pm}=\frac{-\left(\hat{y}_{1}+b \hat{y}_{2}\right) \pm \sqrt{\left(\hat{y}_{1}+b \hat{y}_{2}\right)^{2}-4 b(1-a c) \hat{y}_{1} \hat{y}_{2}}}{2} .
$$

We need to determine conditions under which both $\lambda_{-}$and $\lambda_{+}$will be negative, and since $\lambda_{-} \leq \lambda_{+}$, we only need to check $\lambda_{+}$. For this, notice that $\lambda_{+}$will be negative as long as

$$
4 b(1-a c) \hat{y}_{1} \hat{y}_{2}>0
$$

because in that case the radical will be smaller than $\left(\hat{y}_{1}+b \hat{y}_{2}\right)$.

## Stability

We see that our condition for stability is simply

$$
1-a c>0
$$

In our original coordinates, this is

$$
1-\frac{s_{1} K_{2}}{K_{1}} \frac{s_{2} K_{1}}{K_{2}}=1-s_{1} s_{2}>0 \Longrightarrow s_{1} s_{2}<1
$$

This is precisely the condition already discussed above in the context of the signs of the populations associated with this equilibrium point. Notice that the instability of this equilibrium point in the case $s_{1} s_{2}>1$ is why it was described above as uninteresting.

On this next slide, we'll summarize these observations.

## Stability

We found the following:

- $\hat{y}=\binom{0}{0}$ is always unstable;
- If $s_{2} K_{1}>K_{2}$, then $\hat{y}=\binom{1}{0}$ is asymptotically stable;
- If $s_{1} K_{2}>K_{1}$, then $\hat{y}=\binom{0}{1}$ is asymptotically stable;
- If $s_{2} K_{1}<K_{2}, s_{1} K_{2}<K_{1}$, and $1-s_{1} s_{2}>0$, then there is an equilibrium point $\hat{y}=\binom{\frac{1-a}{1-a c}}{\frac{1-c}{1-a c}}$ with two positive populations, and it is asymptotically stable.

