Interspecific interactions → Predation

Predator ← specialist / generalist

Prey

Lotka Volterra Model

$$\frac{dN}{dt} = rN - aNP$$

$$\frac{dP}{dt} = baNP - mP$$

r ~ prey growth rate

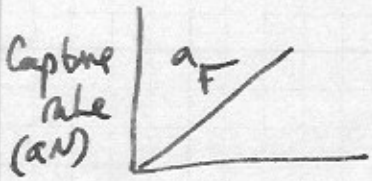
a ~ capture efficiency of prey

aN = # of prey killed per unit time (capture rate)

b ~ efficiency that prey mass is turned over to predator mass

m ~ Natural mortality of predators

* Linear functional response



↳ The more prey, the greater rate of capture

Find E.P.!

$rN = aNP \rightarrow r = aP \rightarrow P^* = \frac{r}{a}$ } when does $P^* \uparrow, \downarrow$

$baNP = mP$

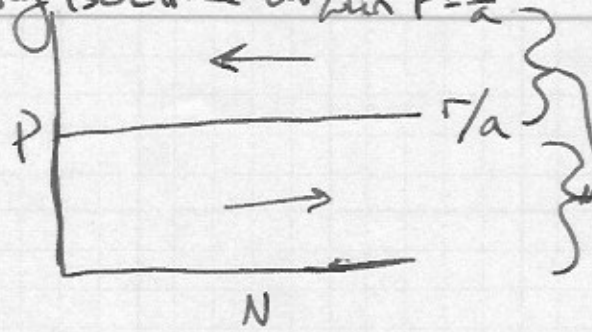
↳ $baN\left(\frac{r}{a}\right) = m\left(\frac{r}{a}\right)$

$N^* = \frac{m}{ba}$ } When does $N^* \uparrow, \downarrow$?

Now we have 2 isoclines: one for prey, one for pred

8.2

Prey Isocline $\frac{dN}{dt} = 0$ when $P = \frac{r}{a}$



Decline

$\frac{dN}{dt} < 0$ when $rN - aNP < 0$

$rN < aNP$

$r < aP$

$\frac{r}{a} < P$ (or) $P > \frac{r}{a}$

Growth

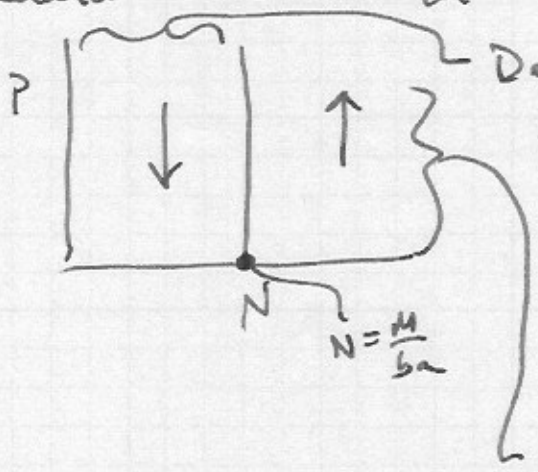
$\frac{dN}{dt} > 0$ when $rN - aNP > 0$

$rN > aNP$

$r > aP$

$r/a > P$ (or) $P < \frac{r}{a}$

Predator Isocline $\frac{dP}{dt} = 0$ when $N = \frac{m}{ba}$



Decline

$\frac{dP}{dt} < 0$ when $baNP - mP < 0$

$baNP < mP$

$baN < m$

$N < \frac{m}{ba}$

Growth

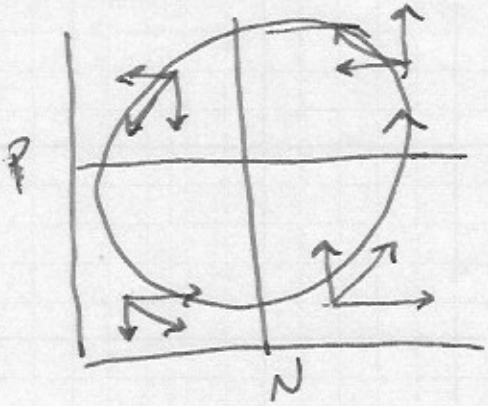
$\frac{dP}{dt} > 0$ when $baNP - mP > 0$

$baNP > mP$

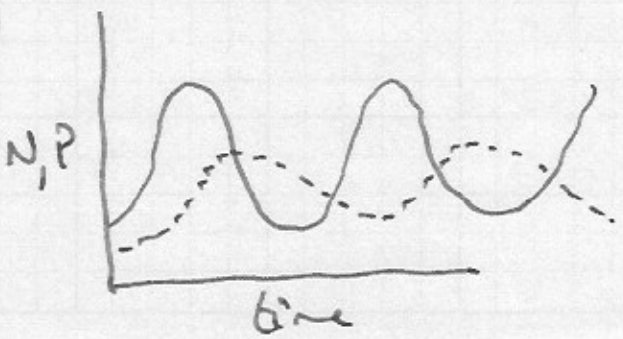
$baN > m$

$N > \frac{m}{ba}$

Combine Isoclines to uncover dynamics



} Predator-prey cycles!

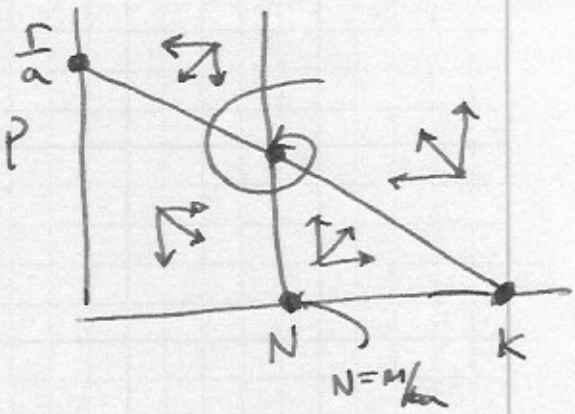


* Diffent initial condition leads to ~~diffent~~ cycles w/ diffent amplitudes

Do the same thing for

$$\frac{dN}{dt} = rN\left(1 - \frac{N}{k}\right) - aNP$$

$$\frac{dP}{dt} = baNP - mP$$



Density-dependent regulation of prey suppresses the cycling.

— How can we learn about the NATURE of a fixed point? Attractor, Repellor, Cycles?

Given $\dot{x} = f(x, y)$, $\dot{y} = g(x, y)$ we first solve (x^*, y^*)

Next we study the dynamics around (x^*, y^*)

Specifically, do perturbations grow or decay

Define perturbations:

$$\epsilon = x - x^*$$
$$\eta = y - y^*$$

$$\dot{\epsilon} = \dot{x} = f(x, y) = \overbrace{f(x^* + \epsilon, y^* + \eta)}^{\text{Taylor Expand}}$$

$$\dot{\epsilon} = f(x^*, y^*) + \epsilon \left.\frac{\partial f}{\partial x}\right|_{(x^*, y^*)} + \eta \left.\frac{\partial f}{\partial y}\right|_{(x^*, y^*)} + \text{h.o.t.}$$

$$\dot{\epsilon} = \epsilon \left.\frac{\partial f}{\partial x}\right|_* + \eta \left.\frac{\partial f}{\partial y}\right|_* + \text{h.o.t.}$$

Stuff

We have:

$$\dot{\epsilon} = \epsilon \left.\frac{\partial f}{\partial x}\right|_* + \eta \left.\frac{\partial f}{\partial y}\right|_* + \text{h.o.t.}$$

Similarly,

$$\dot{\eta} = g(x^*+\epsilon, y^*+\eta) = g(x^*, y^*) + \epsilon \left.\frac{\partial g}{\partial x}\right|_* + \eta \left.\frac{\partial g}{\partial y}\right|_* + \text{h.o.t.}$$

$$\dot{\eta} = \epsilon \left.\frac{\partial g}{\partial x}\right|_* + \eta \left.\frac{\partial g}{\partial y}\right|_* + \text{h.o.t.}$$

We've linearized the problem around the fixed point.

In matrix notation:

$$\begin{pmatrix} \dot{\epsilon} \\ \dot{\eta} \end{pmatrix} = \underbrace{\begin{pmatrix} \left.\frac{\partial f}{\partial x}\right|_* & \left.\frac{\partial f}{\partial y}\right|_* \\ \left.\frac{\partial g}{\partial x}\right|_* & \left.\frac{\partial g}{\partial y}\right|_* \end{pmatrix}}_{\text{Jacobian Matrix } J} \begin{pmatrix} \epsilon \\ \eta \end{pmatrix}$$

Ex) given $\dot{x} = -x + x^3$, $\dot{y} = -2y$

a) find all F.P.

b) Linearize around fixed points given $\begin{pmatrix} \dot{\epsilon} \\ \dot{\eta} \end{pmatrix} = J \begin{pmatrix} \epsilon \\ \eta \end{pmatrix}$

$f(x) = -x + x^3 = 0 \Rightarrow x(x^2-1) = 0$

$g(y) = -2y = 0$

$\hookrightarrow x(x-1)(x+1) = 0$

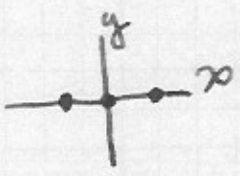
$\begin{cases} x^* = 1, -1, 0 \\ y^* = 0 \end{cases}$

F.P.

$(1, 0)$

$(-1, 0)$

$(0, 0)$



8.5

(b) Linearize the system:

$$J = \begin{pmatrix} \frac{\partial \dot{x}}{\partial x} & \frac{\partial \dot{x}}{\partial y} \\ \frac{\partial \dot{y}}{\partial x} & \frac{\partial \dot{y}}{\partial y} \end{pmatrix} = \begin{pmatrix} -1+3x^2 & 0 \\ 0 & -2 \end{pmatrix}$$

Now - examine near F.P.

Near $(1,0)$ $\quad \begin{pmatrix} \dot{\epsilon} \\ \dot{\eta} \end{pmatrix} = \begin{pmatrix} 2 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} \epsilon \\ \eta \end{pmatrix}$

Near $(0,0)$ $\quad \begin{pmatrix} \dot{\epsilon} \\ \dot{\eta} \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} \epsilon \\ \eta \end{pmatrix}$

Near $(-1,0)$ $\quad \begin{pmatrix} \dot{\epsilon} \\ \dot{\eta} \end{pmatrix} = \begin{pmatrix} 2 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} \epsilon \\ \eta \end{pmatrix}$

(c) Solutions to Linear Systems

In 1D we had $\dot{\eta} = \frac{\partial \dot{x}}{\partial x}\eta$, and solution was

$\eta(t) = \eta_0 e^{\lambda t}$ where $\lambda = \frac{\partial \dot{x}}{\partial x}$

In 2-D (really, n-D), we have the general linear form

$\vec{\dot{\eta}} = J\vec{\eta}$ and we seek solutions of the form:

$\vec{\eta} = \vec{v} e^{\lambda t}$ $\quad \vec{v}$ = eigenvector

$\boxed{\lambda = \text{eigenvalue}}$ — we will focus on this

$\vec{\dot{\eta}} = \vec{v}\lambda e^{\lambda t}$

(get 4.6)

so: $\vec{\dot{\eta}} = \lambda \vec{\eta}$ b/c $\vec{\eta} = \vec{v} e^{\lambda t}$

and: $\vec{\dot{\eta}} = \lambda \vec{\eta} = J\vec{\eta} = J\vec{v}e^{\lambda t}$

so: J

8.6

We seek solutions in form
$\eta(t) = e^{\lambda t}\vec{v} \rightarrow \dot{\vec{\eta}} = \lambda e^{\lambda t}\vec{v}$

We currently have $\dot{\vec{\eta}} = J\eta \rightarrow J\vec{v}e^{\lambda t}$

so: $\lambda e^{\lambda t}\vec{v} = e^{\lambda t} J \vec{v}$

$J\vec{v} = \lambda\vec{v}$ } If $\vec{v}$ is an eigenvector of J, and with corresponding eigenvalue λ, then linear solutions exist.

Focus on eigenvalues...

KEY: The eigenvalues of $J|_*$ will give us the growth rates of the linear approximation to dynamics around the Fixed point ($*$).

How do you find the eigenvalues of J?

Is given by the characteristic equation: $\det(J - \lambda I) =$

What is $J - \lambda I$? if $J = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$, $I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$

and $J - \lambda I = \begin{pmatrix} a-\lambda & b \\ c & d-\lambda \end{pmatrix}$

What is $\det(\cdot)$? for matrix $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$, det. is $ad - bc$

so: $\det(J - \lambda I) = (a-\lambda)(d-\lambda) - bc$

Another characteristic of matrices is Trace: $a + d$ for 2x2 matrix

8.7

$\tau = \text{trace}(\vec{\vec{J}}) = a + d$

$\Delta = \det(\vec{\vec{J}}) = ad - bc$

then $\lambda_1 = \frac{\tau + \sqrt{\tau^2 - 4\Delta}}{2}$, $\lambda_2 = \frac{\tau - \sqrt{\tau^2 - 4\Delta}}{2}$

- Notice there are 2 eigenvalues for the 2-D systems and they depend only on the trace and determinant of $\vec{\vec{J}}$!

- Eigenvalues are distinct (generally), and eigenvectors are linearly independent, which allows us to write general solution for $\vec{\eta}(t)$ as

$$\vec{\eta}(t) = c_1 e^{\lambda_1 t} \vec{v}_1 + c_2 e^{\lambda_2 t} \vec{v}_2$$ where c_1, c_2 are constants.

Consider ~~the logistic~~ L-V predator system:

$$\frac{dN}{dt} = rN - aNP = f(N, P)$$

$$\frac{dP}{dt} = baNP - mP = g(N, P)$$

There are 2 fixed points: $(0, 0)$

$\left(\frac{m}{ba}, \frac{r}{a}\right)$

What is the Jacobian?

$$J = \begin{pmatrix} \frac{\partial f}{\partial N} & \frac{\partial f}{\partial P} \\ \frac{\partial g}{\partial N} & \frac{\partial g}{\partial P} \end{pmatrix}\Bigg|_* = \begin{pmatrix} r - aP & -aN \\ baP & baN - m \end{pmatrix}$$

8.8

$$\tilde{J} = \begin{pmatrix} r - aP & -aN \\ baP & baN - m \end{pmatrix}$$

$$\tilde{J}\Big|_{(0,0)} = \begin{pmatrix} r & 0 \\ 0 & -m \end{pmatrix}$$

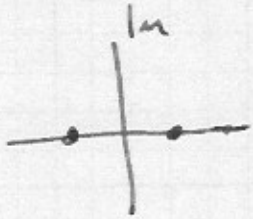
$$\tilde{J}\Big|_{\left(\frac{m}{ba}, \frac{r}{a}\right)} = \begin{pmatrix} r - a\left(\frac{r}{a}\right) & -a\left(\frac{m}{ba}\right) \\ ba\left(\frac{r}{a}\right) & ba\left(\frac{m}{ba}\right) - m \end{pmatrix}$$

$$\tilde{J}\Big|_{\left(\frac{m}{ba}, \frac{r}{a}\right)} = \begin{pmatrix} 0 & -\frac{m}{b} \\ br & 0 \end{pmatrix}$$

Find Eigenvalues

$$\lambda_{1,2}\Big|_{(0,0)} = \frac{\tau \pm \sqrt{\tau^2 - 4\Delta}}{2} = \frac{1}{2}\left(r - m \pm \sqrt{(m+r)^2}\right)$$

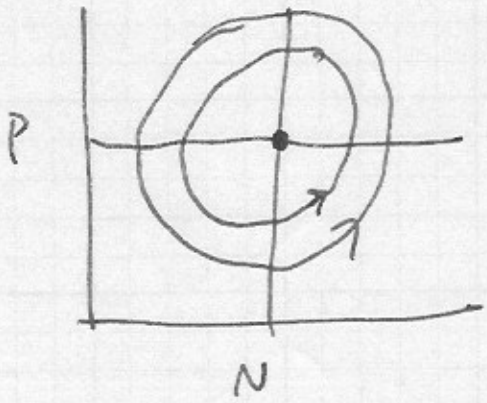
High Positive branch: $\lambda_1 > 0$
Low branch: $\lambda_2 < 0$



$$\lambda_{1,2}\Big|_{\left(\frac{m}{ba}, \frac{r}{a}\right)} = \pm\sqrt{-mr} = \pm i\sqrt{mr}$$ } eigenvalues purely imaginary

$$\begin{cases} \Delta = mr \\ \tau = 0 \end{cases}$$

From graphical analysis:



(N^*, P^*) is a center

"Centers"

- any initial condition leads to a unique cycle

In 2-D, this is always true:

$$\lambda_{1,2} = \frac{\tau \pm \sqrt{\tau^2 - 4\Delta}}{2}$$

$\Delta = \lambda_1 \lambda_2$

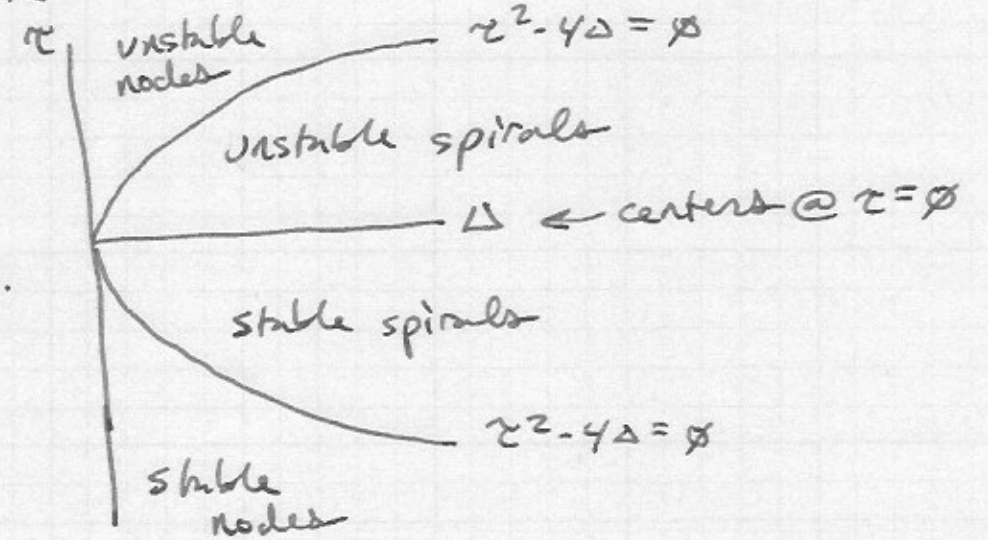
$\tau = \lambda_1 + \lambda_2$

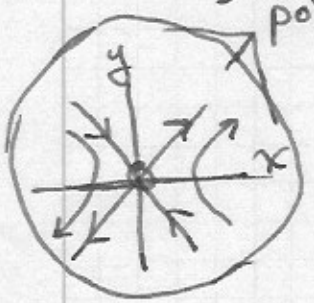
- if one eigenvalue is (+), and one is negative, Δ is (-)
- if both eigs are (+,+) or (-,-), Δ is (+)

↳ Stability requires that Δ is (+), but is not a sufficient condition to guarantee it.

- When $\tau < 0$, both eigenvalues have negative real parts, so fixed point is STABLE if $\Delta > 0$
- When $\tau > 0$, and $\Delta > 0$, both eigenvalues have positive real parts ∴ UNSTABLE
- if $\tau^2 - 4\Delta < 0$, there are imaginary parts to eigenvalues ↳ cycles

PUT TOGETHER





(illustration of saddle point)