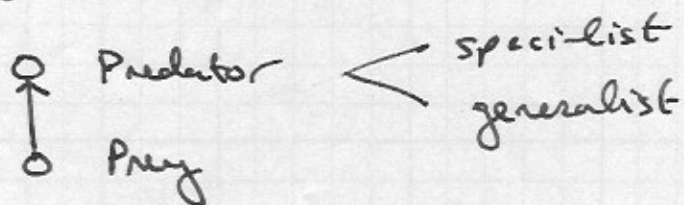


Intraspecific Interactions \rightarrow Predation

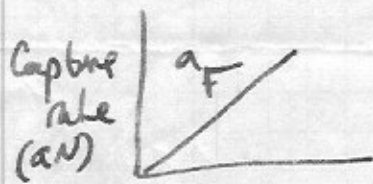
Lotka Volterra Model

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

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

 $r \sim$ prey growth rate $a \sim$ capture efficiency of prey $aN = \#$ of prey killed per unit time (capture rate) $ba \sim$ efficiency that prey mass is turned over to predator mass $m \sim$ Natural mortality of predators

① Linear functional response



The more prey, the greater rate of capture

Find F.P.!

$$rN = aNP \rightarrow r = aP \rightarrow P^* = \frac{r}{a} \left\{ \begin{array}{l} \text{when does } P^* \uparrow, \downarrow \end{array} \right.$$

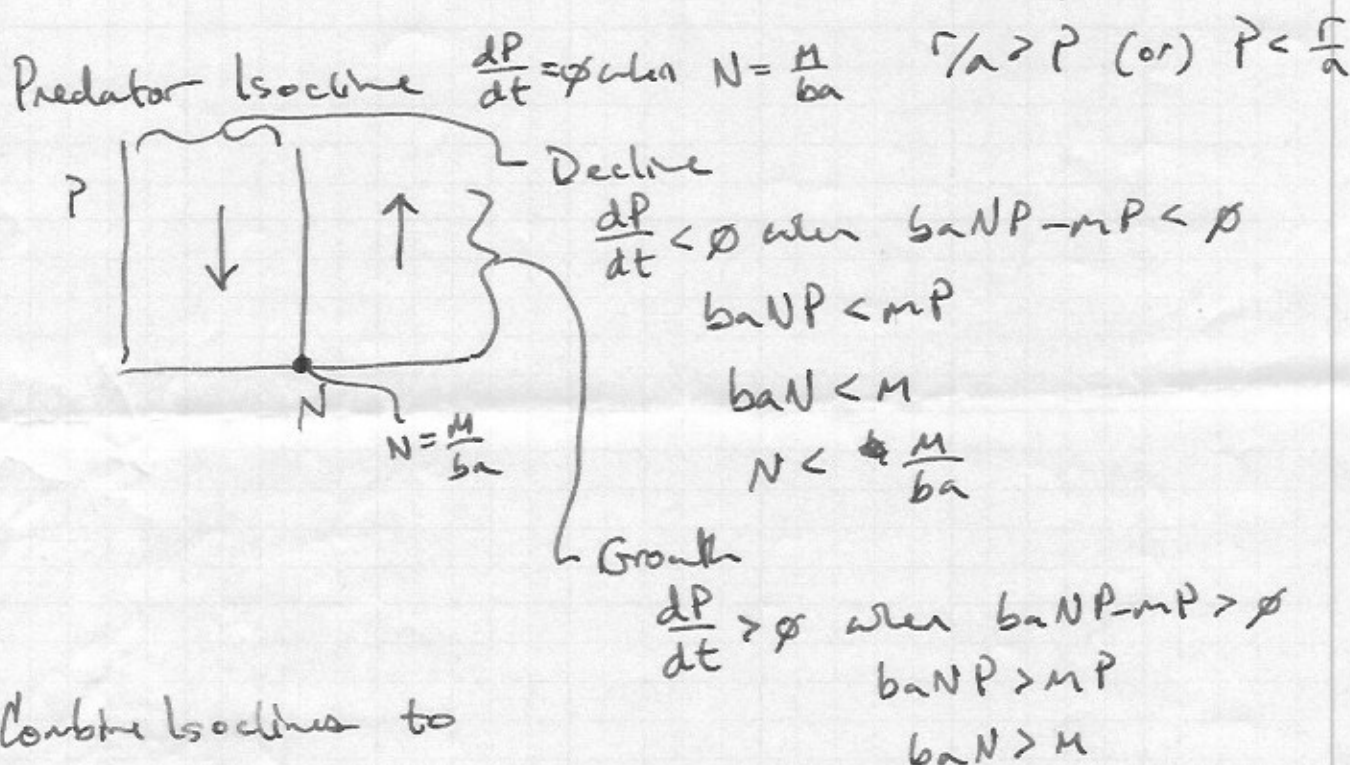
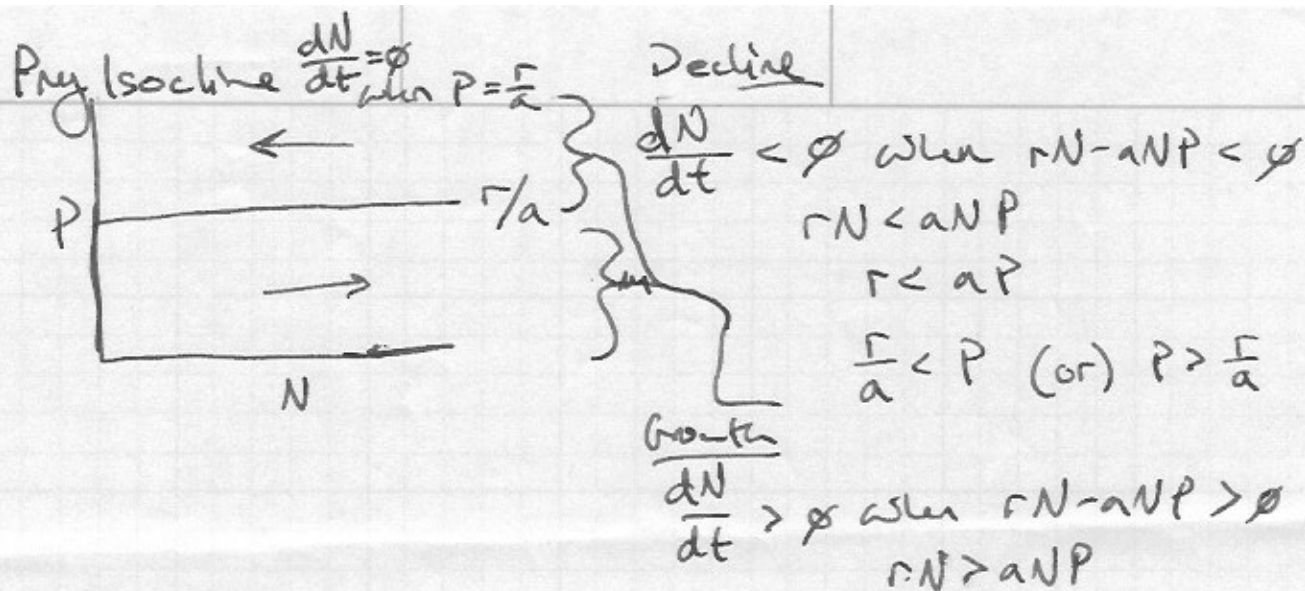
$$baNP = mP$$

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

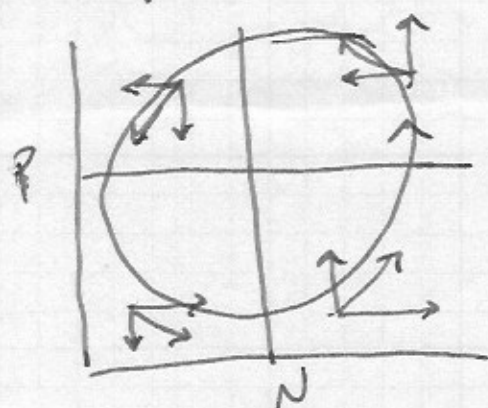
$$N^* = \frac{m}{ba} \left\{ \begin{array}{l} \text{when does } N^* \uparrow, \downarrow \end{array} \right.$$

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

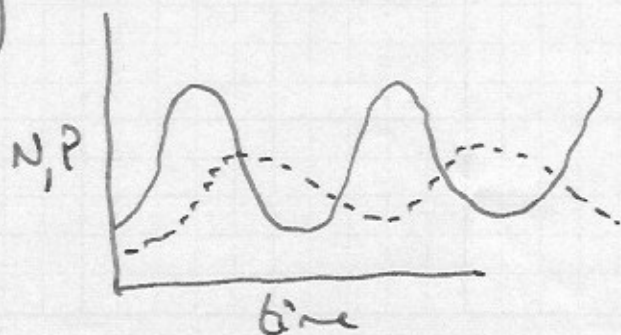
8.2



Combine Isoclines to
uncover dynamics



Predator-prey cycles!

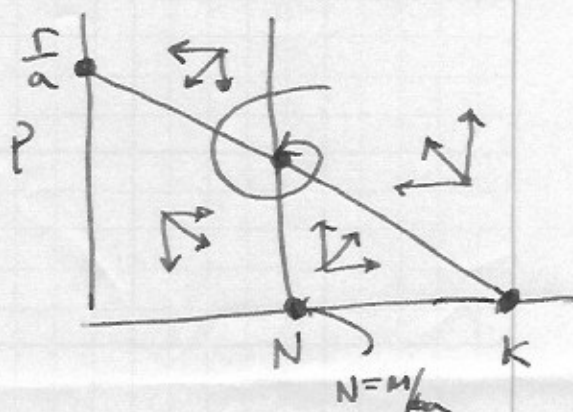


⊛ Different initial condition
leads to ~~different~~ cycles w/
different amplitudes

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 $x = f(x, y)$
 $y = g(x, y)$ we first solve (x^*, y^*)

Next we study the dynamics around (x^*, y^*)
 Specifically, do perturbations grow or decay

Define perturbations:

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

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

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

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

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

Start

We have:

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

Similarly,

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

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

⊗ We've linearized the problem around the fixed point.
In matrix notation:

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

Jacobian Matrix

Ex) given $\dot{x} = -x + x^3$ a) find all F.P.
 $\dot{y} = -2y$ 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:

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

Now - examine near F.P.

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

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

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

(c) Solutions to Linear Systems

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

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

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

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

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

 ~~$\vec{v} = \text{eigenvector}$~~ ~~$\lambda = \text{eigenvalue}$ - we will focus on this~~

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

(see 8.6)

~~so: $\dot{\vec{\eta}} = \lambda \vec{\eta}$~~

~~$\dot{\vec{\eta}} = \lambda \vec{\eta} \quad \text{b/c } \vec{\eta} = \vec{v} e^{\lambda t}$~~

~~substitute into~~

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

~~so: $J\vec{v} = \lambda \vec{v}$~~

8.6

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

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

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

~~$\lambda e^{\lambda t} \vec{v}$~~

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

Focus on eigenvalues...

KEY

The eigenvalues of J_* will give us the growth rates of the ~~at~~ linear approximation to dynamics around the Fixed point (*).

How do you find the eigenvalues of J ?

↳ given by the characteristic equation: $\det(J - \lambda I) = 0$

~~$\det(J - \lambda I)$~~

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 2×2 matrix

$$\tau = \text{trace}(\mathcal{A}) = a + d$$

$$\Delta = \det(\mathcal{A}) = ad - bc$$

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

- Notice there are 2 eigenvalues for the 2-D system, and they depend only on the trace and determinant of \mathcal{A} !

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

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

Consider the ~~the~~ ~~the~~ ~~the~~

L-V pred-prey system:

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

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

There are 2 fixed points: $(0, 0)$
 $(\frac{m}{ba}, \frac{r}{a})$

What is the Jacobian?

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

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

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

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

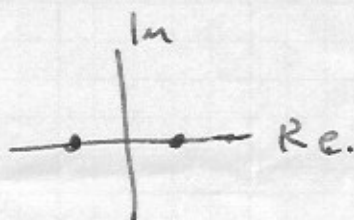
$$J|_{\left(\frac{m}{ba}, \frac{r}{a}\right)} = \begin{pmatrix} 0 & -\frac{m}{b} \\ br & 0 \end{pmatrix}$$

Find Eigenvalues

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

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

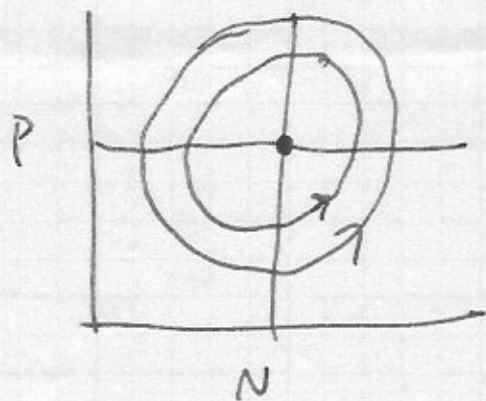
~~asymptotically~~
stable



$$\lambda_{1,2}|_{\left(\frac{m}{ba}, \frac{r}{a}\right)} = \pm \sqrt{-mr} = \pm i\sqrt{mr} \quad \left. \vphantom{\lambda_{1,2}} \right\} \begin{array}{l} \text{eigenvalues} \\ \text{purely imaginary} \end{array}$$

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

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 (+)

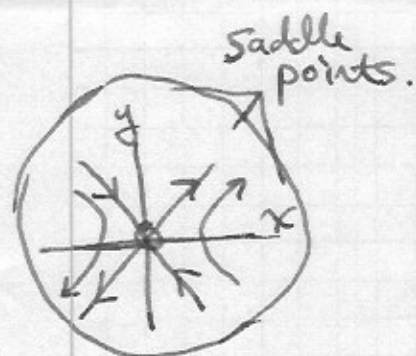
↳ ~~stability~~ stability requires that Δ is (+), but is not a sufficient condition to ~~guarantee~~ guarantee it.

- 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 ~~are~~ have positive real parts ~ UNSTABLE

- if $\tau^2 - 4\Delta < 0$, there are imaginary parts to eigenvalues
↳ cycles

POT TOGETHER



(illustration of saddle point)

