Nonlinear Systems

Chapter 2 and 3 Read Them!!

Nonlinear Systems

$$x_{n+1} = f(x_n, y_n)$$
$$y_{n+1} = g(x_n, y_n)$$

- Notes:
 - The functions f and g depend nonlinearly on their arguments
 - Usually these systems cannot reduce to a single equation

Nonlinear Systems $x_{n+1} = f(x_n, y_n)$ $y_{n+1} = g(x_n, y_n)$

Steady States:

$$x_e = f(x_e, y_e)$$
$$y_e = g(x_e, y_e)$$

- A system of 2 equations and 2 unknowns must be solved in order to determine the steady states.
- Sometimes it will difficult or impossible to solve for steady states analytically.

$$x_{n+1} = f(x_n, y_n)$$
 $y_{n+1} = g(x_n, y_n)$

 Follow the same steps that were outlined for a single nonlinear equation.

1. Let

$$\begin{aligned} x_n &= x_e + \overline{x}_n \\ y_n &= y_e + \overline{y}_n \end{aligned} \qquad \left| \overline{x}_n, \overline{y}_n \right| << 1 \end{aligned}$$

$$x_{n+1} = f(x_n, y_n)$$
 $y_{n+1} = g(x_n, y_n)$

2. Substitute into the model equations:

$$\begin{aligned} x_n &= x_e + \overline{x}_n \\ y_n &= y_e + \overline{y}_n \end{aligned}$$
$$\begin{aligned} x_e + \overline{x}_{n+1} &= f(x_e + \overline{x}_n, y_e + \overline{y}_n) \\ y_e + \overline{y}_{n+1} &= g(x_e + \overline{x}_n, y_e + \overline{y}_n) \end{aligned}$$

3. Expand the right-hand side of each equation in a Taylor series:

$$f(x_e + \bar{x}_n, y_e + \bar{y}_n) = f(x_e, y_e) + \frac{\partial f}{\partial x}(x_e, y_e)\bar{x}_n + \frac{\partial f}{\partial y}(x_e, y_e)\bar{y}_n + \dots + O(\bar{x}_n^2, \bar{y}_n^2)$$

$$g(x_e + \overline{x}_n, y_e + \overline{y}_n) = g(x_e, y_e) + \frac{\partial g}{\partial x}(x_e, y_e)\overline{x}_n + \frac{\partial g}{\partial y}(x_e, y_e)\overline{y}_n + \dots + O(\overline{x}_n^2, \overline{y}_n^2)$$

4. Neglect higher order terms:

$$f(x_e + \bar{x}_n, y_e + \bar{y}_n) = f(x_e, y_e) + \frac{\partial f}{\partial x}(x_e, y_e)\bar{x}_n + \frac{\partial f}{\partial y}(x_e, y_e)\bar{y}_n$$
$$g(x_e + \bar{x}_n, y_e + \bar{y}_n) = g(x_e, y_e) + \frac{\partial g}{\partial x}(x_e, y_e)\bar{x}_n + \frac{\partial g}{\partial y}(x_e, y_e)\bar{y}_n$$

5. Simplify



Let

$$a_{11} = \frac{\partial f}{\partial x}(x_e, y_e)$$
 $a_{12} = \frac{\partial f}{\partial y}(x_e, y_e)$

$$a_{21} = \frac{\partial g}{\partial x}(x_e, y_e)$$
 $a_{22} = \frac{\partial g}{\partial y}(x_e, y_e)$

5. After Simplification

$$\overline{x}_{n+1} = a_{11}\overline{x}_n + a_{12}\overline{y}_n$$
$$\overline{y}_{n+1} = a_{21}\overline{x}_n + a_{22}\overline{y}_n$$

Therefore, a linear system governs the behavior of the perturbations.

6. Reduce to a single higher order equation to find characteristic equation:

$$\overline{x}_{n+2} - (a_{11} + a_{22})\overline{x}_{n+1} + (a_{12}a_{21} - a_{22}a_{11})\overline{x}_n = 0$$

$$\lambda^2 - (a_{11} + a_{22})\lambda + (a_{12}a_{21} - a_{22}a_{11}) = 0$$

$$\lambda_{1,2} = \frac{a_{11} + a_{22} \pm \sqrt{(a_{11} + a_{22})^2 - 4(a_{11}a_{22} - a_{12}a_{21})}}{2}$$

7. Determine the magnitude of the eigenvalues:

Let
$$\beta = a_{11} + a_{22}$$
 $\gamma = a_{11}a_{22} - a_{12}a_{21}$
 $\lambda_{1,2} = \frac{\beta \pm \sqrt{\beta^2 - 4\gamma}}{2}$

It turns out that $|\lambda_{1,2}| < 1$ if $|\beta| < 1 + \gamma < 2$

Stability Conclusion

The steady states of the nonlinear system:

$$x_{n+1} = f(x_n, y_n)$$
 $y_{n+1} = g(x_n, y_n)$

• Are stable if
$$\left| \lambda_{1,2} = \frac{\beta \pm \sqrt{\beta^2 - 4\gamma}}{2} \right| < 1$$

• The reduces to $|\beta| < 1 + \gamma < 2$

Ecological Example

Host-Parasitoid Systems

Parasitoids

- Definition: Insects that have an immature life stage that develops on or within a single insect host, ultimately killing the host.
- Major Characteristics
 - they are specialized in their choice of host
 - they are smaller than host (a few mm long, usually)
 - only the female searches for host
 - different parasitoid species can attack different life stages of host
 - eggs or larvae are usually laid in, on, or near host
 - immatures remain on or in host and almost always kill host
 - adults are free-living, mobile, and may be predaceous

Parasitoid Life Cycle



This often follows an annual cycle.

Parasitoid Facts

- Parasitoids are an incredibly diverse and successful type of <u>insect</u>.
- There are 50,000 and 15,000 described species of parasitoid wasps and flies respectively, along with around 3,000 in other orders.
- Parasitoids make up about 8.5% of all insect species.
- They are used biological agents for the control of insect pests.

Let's Try To Model This Behavior!

Model Assumptions

- Hosts that have been infected in the previous generation will give rise to the next generation parasitoids.
- Hosts that have not been infected give rise to the next generation of hosts.
- Fraction of hosts that are infected depends on the "searching efficiency" of the parasitoid population or "contact rate" of the two populations.

Variables and Parameters

- N_t = density of host population in generation t
- P_t = density of parasitoid population in generation t
- φ(N_t, P_t) = fraction of hosts not infected in generation t, ie the probability of escaping parasitism
- λ = host reproductive rate
- c = average number of viable eggs laid by a parasitoid in a single host

Building the Model

N _{t+1} =	# of hosts in previous generation	*	probability of hosts not being infected	*	reproduction rate
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P _{t+1} =	# of hosts in previous generation	*	probability of hosts being infected	*	# of eggs produced per host
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General Model Equations

$$N_{t+1} = \lambda N_t \phi(N_t, P_t)$$

$$P_{t+1} = cN_t \left[1 - \phi(N_t, P_t) \right]$$

Nicholson-Bailey Model

- 1. Parasitoids search independently and encounters occur randomly
- 2. The searching efficiency is constant
- Only the first encounter between the host and the parasitoid is significant.
 The probability of escaping parasitism is the same as the probability of zero encounters:

$$\phi(N_t, P_t) = e^{-aP_t}$$

Nicholson-Bailey Model Equations

$$N_{t+1} = \lambda N_t e^{-aP_t}$$
$$P_{t+1} = c N_t \left[1 - e^{-aP_t} \right]$$

What happens to the hosts when there are no parasitoids? What happens to the parasitoids if there are no hosts?

• Steady States

- Let
$$N_{t+1} = N_t = N_e$$
 $P_{t+1} = P_t = P_e$

- Substitute into Model Equations

$$N_e = \lambda N_e e^{-aP_e}$$
$$P_e = cN_e \left[1 - e^{-aP_e}\right]$$

- Solve

$$N_{e1} = 0$$
 $N_{e2} = \frac{\lambda \ln \lambda}{ac(\lambda - 1)}$ $P_{e2} = \frac{\ln \lambda}{a}$

- Steady State Summary
 - Two sets of steady states
 - The elimination state:

$$N_{e1} = 0 \qquad P_{e1} = 0$$

• The coexistence state:



Stability

-Let
$$f(N,P) = \lambda N e^{-aP}$$
 $g(N,P) = cN(1-e^{-aP})$

- Compute

$$a_{11} = \frac{\partial f}{\partial N}(N_e, P_e) \qquad a_{12} = \frac{\partial f}{\partial P}(N_e, P_e)$$
$$a_{21} = \frac{\partial g}{\partial N}(N_e, P_e) \qquad a_{22} = \frac{\partial g}{\partial P}(N_e, P_e)$$

• Stability of the Elimination State $f(N,P) = \lambda N e^{-aP}$ $g(N,P) = cN(1-e^{-aP})$

$$a_{11} = \frac{\partial f}{\partial N}(0,0) = \lambda e^{-a(0)} = \lambda \qquad \qquad a_{12} = \frac{\partial f}{\partial P}(0,0) = -a\lambda(0)e^{-a(0)} = 0$$

$$a_{21} = \frac{\partial g}{\partial N}(0,0) = c\left(1 - e^{-a(0)}\right) = 0 \qquad a_{22} = \frac{\partial g}{\partial P}(0,0) = -ac(0)e^{-a(0)} = 0$$

• Stability of the Elimination State $f(N,P) = \lambda N e^{-aP}$ $g(N,P) = cN(1 - e^{-aP})$

$$\beta = a_{11} + a_{22} = \lambda \qquad \gamma = a_{11}a_{22} - a_{12}a_{21} = 0$$

Stability Condition: $|\beta| < 1 + \gamma < 2$

$$\lambda < 1$$

The elimination state is stable if and only if the compromise state DNE

Stability of the Compromise State

$$N_{e2} = \frac{\lambda \ln \lambda}{ac(\lambda - 1)}$$
 $P_{e2} = \frac{\ln \lambda}{a}$

$$\begin{aligned} a_{11} &= \frac{\partial f}{\partial N} (N_{e2}, P_{e2}) = \lambda e^{-a(P_{e2})} = 1 \\ a_{12} &= \frac{\partial f}{\partial P} (N_{e2}, P_{e2}) = -a\lambda N_{e2} e^{-a(P_{e2})} = \frac{-\lambda \ln \lambda}{c(\lambda - 1)} \\ a_{21} &= \frac{\partial g}{\partial N} (N_{e2}, P_{e2}) = c\left(1 - e^{-a(P_{e2})}\right) = c\left(1 - \frac{1}{\lambda}\right) \\ a_{22} &= \frac{\partial g}{\partial P} (N_{e2}, P_{e2}) = -acN_{e2} e^{-a(P_{e2})} = \frac{\lambda}{\lambda - 1} \end{aligned}$$

Stability of the Compromise State

$$\beta = a_{11} + a_{22} = 1 + \frac{\ln \lambda}{\lambda - 1}$$
$$\gamma = a_{11}a_{22} - a_{12}a_{21} = \frac{\lambda \ln \lambda}{\lambda - 1}$$

Stability Condition: $|\beta| < 1 + \gamma < 2$

Must Show:
$$1 + \frac{\ln \lambda}{\lambda - 1} < 1 + \frac{\lambda \ln \lambda}{\lambda - 1} < 2$$

Stability of the Compromise State



Conclusions

- The Nicholson-Bailey Model has two steady states. The compromise state is never stable.
- This model predicts that when the compromise state exists both populations will undergo growing oscillations.
- Interestingly, the green house whitefly and its parasitoid was shown to have this behavior under very specific lab conditions.
- The model predicts the exact opposite of the desired effect for a biological control agent.

Numerical Simulation



Experiments with the greenhouse whitefly and its parasitoid, provides the closest correspondence (for nearly 20 generations) with the Nicholson-Bailey model. $\lambda = 2$, c = 1, a = 0.068, initial host 24, initial parasite 12.

Long Time Simulation



 Eventually the parasitoid population crashes and the host population explodes.

Modifying the N-B Model

- Surely, natural systems are more stable than this.
- Let try modifying the assumptions that underlie the host population and investigate whether these modifications have a stabilizing effect.

Modifying the N-B Model

- Assume that in the absence of parasitoids, the host population grows to a limited density determined by the environmental carrying capacity.
- How would the model equations change?

$$N_{t+1} = \lambda (N_t) N_t e^{-aP_t}$$
$$P_{t+1} = c N_t \Big[1 - e^{-aP_t} \Big]$$

Modifying the N-B Model

• Let's choose $\lambda(N_t) = e^{r\left(1 - \frac{N_t}{K}\right)}$

$$N_{t+1} = N_t e^{\left[r\left(1 - \frac{N_t}{K}\right) - aP_t\right]}$$
$$P_{t+1} = cN_t \left[1 - e^{-aP_t}\right]$$

$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



For small r, the nontrivial steady state is stable.

The iterates move along a "spiral galaxy".

Numerical Simulation



• Solutions oscillate towards a stable coexistence equilibrium.



What's going on here?

$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



As r increases, the nontrivial steady state becomes unstable.

The iterates move along a stable limit cycle.

$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



As r increases further, the non-trivial steady state remains unstable.

The iterates move along a 5-point cycle.

$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



What's going on here?

$$N_{t+1} = N_t \exp[r(1 - \frac{N_t}{K}) - aP_t]$$
$$P_{t+1} = N_t (1 - \exp(-aP_t))$$



Still larger values of r yield either chaos or cycles of extremely high period.

This chaotic behavior begins to fill a sharply bounded area of phase space.

Other Potentially Stabilizing Modifications

- Heterogeneity of the environment
 - Part of the population may be less exposed and therefore less vulnerable to attack.
 - How could we model this situation?
 - You'll get the chance to explore this in Homework 3.