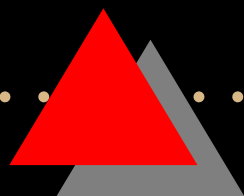


*Ordinary Differential Equations  
and Introduction to Dynamical  
Systems*

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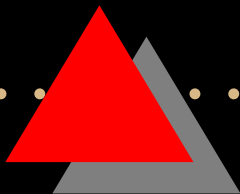
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# Overview

- Single Species Systems
  - Solving for Equilibria
  - Evaluating Stability of Equilibria Graphically
- Two Species Systems
  - Lotka-Volterra Predator-Prey
  - Evaluating Stability of Equilibria
- Examples from Epidemiology



# *Single Species Systems*

- Exponential Growth
- Logistic Growth
- Other Equations

# *Exponential Growth*

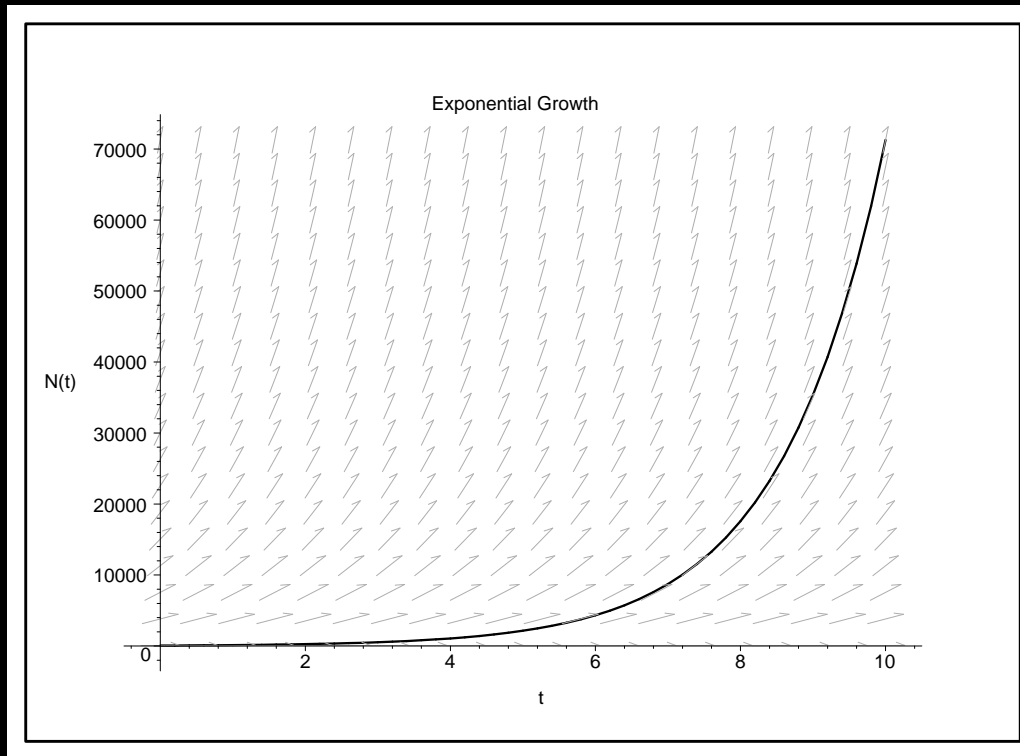
$$\frac{dN}{dt} = rN$$

Solution:

$$N(t) = N_0 e^{rt}$$

What happens to this population?

# Exponential Growth



# Logistic Growth

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

What do you think the solutions of this will look like?

Recall exponential growth was

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

# Equilibria

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

$$\frac{dN}{dt} = 0$$

$$rN^* \left( 1 - \frac{N^*}{K} \right) = 0$$

$$N^* = 0 \quad \text{or} \quad N^* = K$$



# Stability of Equilibria

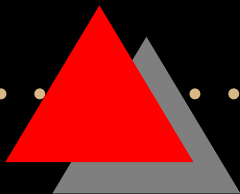
First evaluate the stability of  $N^* = 0$ .

Near  $N^* = 0$ ,

$$\frac{dN}{dt} \approx rN$$

So as  $N$  increase,  $\frac{dN}{dt}$  grows exponentially.

Therefore,  $N^* = 0$  is an unstable equilibrium.



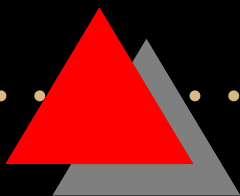




# *Stability of Equilibria*

What do you think will happen near  $N^* = K$ ?

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





# *Stability near $K$*

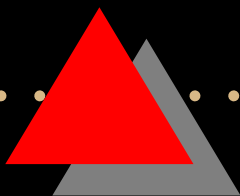
If  $N$  is just slightly above  $K$ ,

$$\frac{dN}{dt} = rN \left( 1 - \frac{N}{K} \right) < 0$$

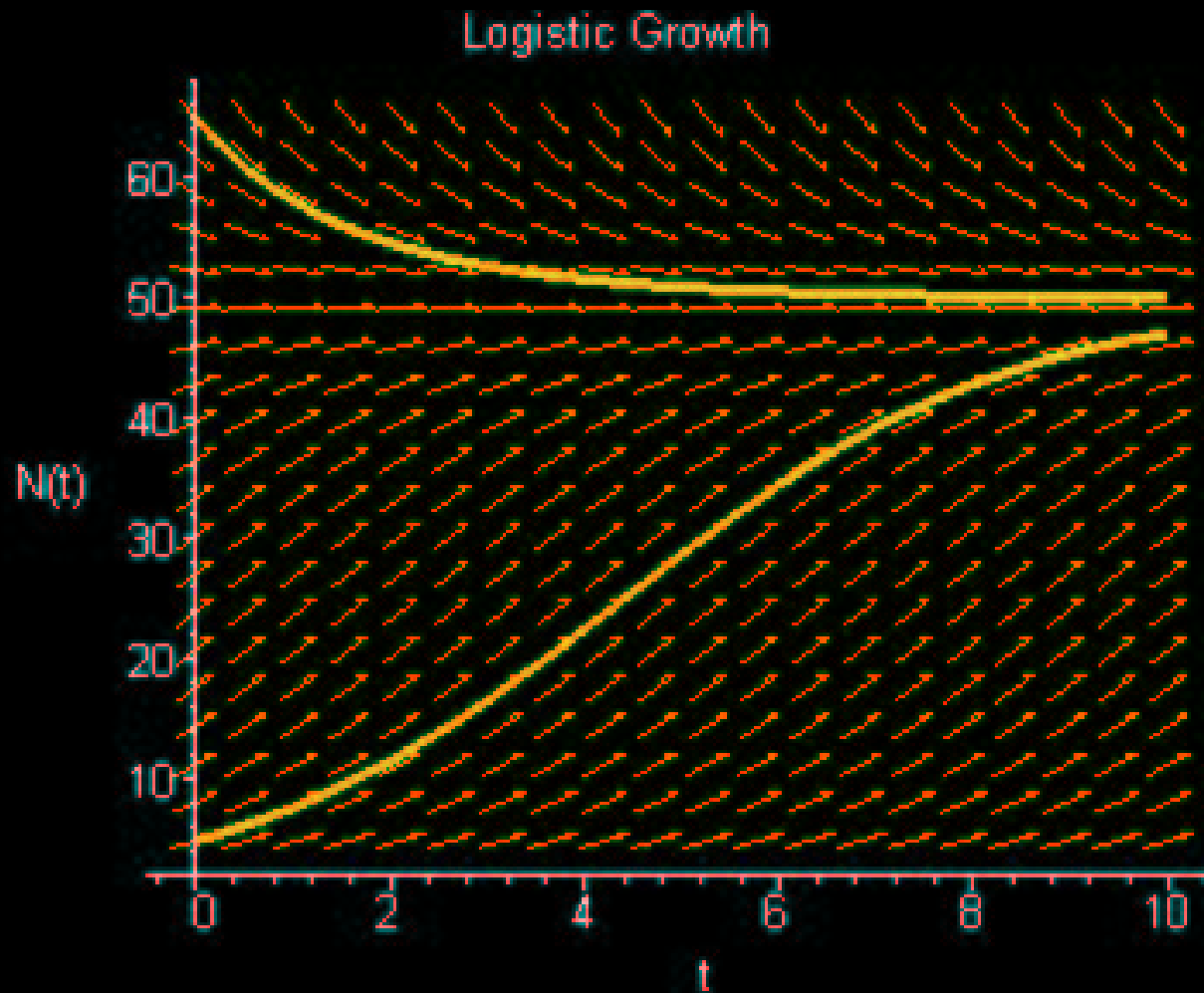
but if  $N$  is just slightly below  $K$ ,

$$\frac{dN}{dt} = rN \left( 1 - \frac{N}{K} \right) > 0$$

Therefore,  $N^* = K$  is stable.



# Graphical View of Stability



# Other Alternatives

- Gompertz Equation

$$\frac{dN}{dt} = r_0 e^{-\alpha t} N$$

- Delay or lag time

$$\frac{dN}{dt} = F(N(t), N(t - T))$$



# Other Alternatives

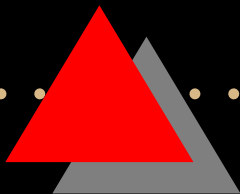
- Allee effect

$$\frac{dN}{dt} = rN \left(1 - \frac{N}{K}\right) \left(\frac{N}{K_0} - 1\right)$$

- Discrete time

$$N(t + 1) = F(N(t))$$

- Stochastic processes

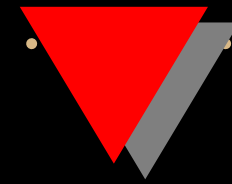




# *Interacting Populations*

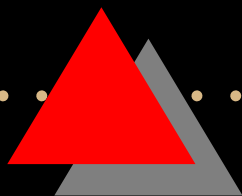
- Predator-prey models
- Competition
- Mutualism

# Classic Predator-Prey



## Lotka-Volterra Predator-Prey Model

$$\frac{dN}{dt} = rN - cNP$$
$$\frac{dP}{dt} = bNP - mP$$

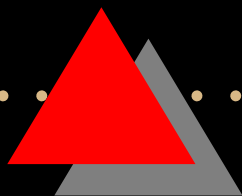




# *Classic Predator-Prey*

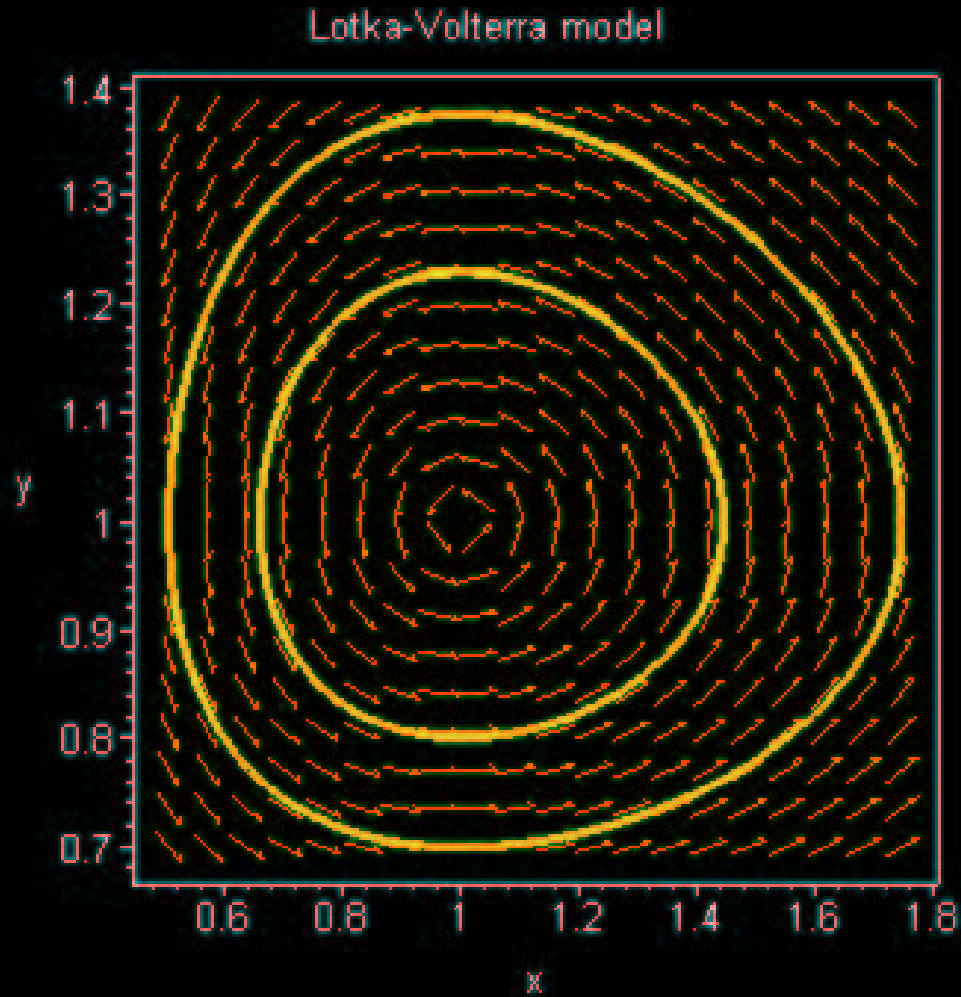
## Lotka-Volterra Predator-Prey Model

- Historical interest
- Mass-action term
- Bad mathematical model
- Structurally unstable





# Lotka-Volterra Phase Plane



# Interacting Populations

More Realistic Predator-prey models

$$\frac{dN}{dt} = rN \left( 1 - \frac{N}{K} \right) - P \left( \frac{A}{N + B} \right)$$

$$\frac{dP}{dt} = eP \left( \frac{A}{N + B} \right) - dP$$

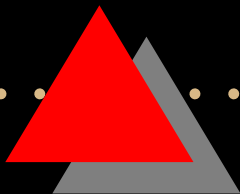


# *Interacting Populations*

Another More Realistic Predator-prey models

$$\frac{dN}{dt} = rN \left( 1 - \frac{N}{K} \right) - P \left( \frac{AN}{N^2 + B^2} \right)$$

$$\frac{dP}{dt} = eP \left( \frac{AN}{N^2 + B^2} \right) - dP$$



# Interacting Populations

## Competition

$$\frac{dN_1}{dt} = r_1 N_1 \left( 1 - \frac{N_1}{K_1} - b_{12} \frac{N_2}{K_1} \right)$$
$$\frac{dN_2}{dt} = r_2 N_2 \left( 1 - \frac{N_2}{K_2} - b_{21} \frac{N_1}{K_2} \right)$$

# Interacting Populations

## Mutualism

$$\frac{dN_1}{dt} = r_1 N_1 \left( 1 - \frac{N_1}{K_1} + b_{12} \frac{N_2}{K_1} \right)$$

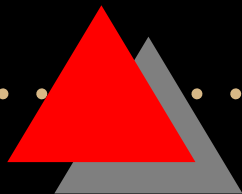
$$\frac{dN_2}{dt} = r_2 N_2 \left( 1 - \frac{N_2}{K_2} + b_{21} \frac{N_1}{K_2} \right)$$



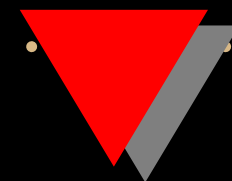
# *Interacting Populations*

To analyze these types of models

- Nondimensionalize the system
  - reduce the number of parameters
  - simplify the system
- Solve for equilibria
- Analyze stability of equilibria
- Translate back to determine biological significance

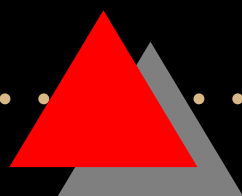


# Phase-Plane Techniques



## Some definitions of stability

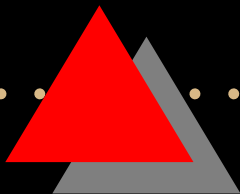
- Stable - if start small distance from equilibrium, remain small distance as  $t \rightarrow \infty$ 
  - Lyapunov stable
  - locally stable
- Asymptotically stable - if start small distance from equilibrium, distance from equilibrium approaches zero as  $t \rightarrow \infty$ 
  - locally asymptotically stable





# *Phase-Plane Techniques*

- Linearization
- Bendixson-Dulac negative criterion
- Hopf bifurcation theorem
- Poincaré-Bendixson theorem
- Routh-Hurwitz Conditions





# Linearization

Given:

$$\frac{dN}{dt} = F(N, P)$$

$$\frac{dP}{dt} = G(N, P)$$

# Linearization

Solve:

$$F(N^*, P^*) = 0$$

$$G(N^*, P^*) = 0$$

to find the equilibria,  $(N^*, P^*)$ .

Let:

$$x(t) = N(t) - N^*$$

$$y(t) = P(t) - P^*$$

# Linearization

Then linearize about the equilibrium:

$$\frac{dx}{dt} = \left. \frac{\partial F}{\partial N} \right|_{(N^*, P^*)} x + \left. \frac{\partial F}{\partial P} \right|_{(N^*, P^*)} y$$

$$\frac{dy}{dt} = \left. \frac{\partial G}{\partial N} \right|_{(N^*, P^*)} x + \left. \frac{\partial G}{\partial P} \right|_{(N^*, P^*)} y$$

Or:

$$\begin{pmatrix} \dot{x} \\ \dot{y} \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

# Linearization

Let:

$$J = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix}$$

Where  $J$  is known as the Jacobian matrix or the community matrix.

We now look for solutions of the form:

$$x(t) = x_0 e^{\lambda t}$$

$$y(t) = y_0 e^{\lambda t}$$



# Linearization

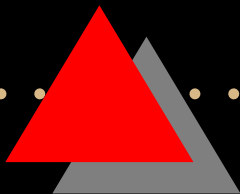
Substitute this back into the equations to obtain:

$$\lambda x_0 = a_{11}x_0 + a_{12}y_0$$

$$\lambda y_0 = a_{21}x_0 + a_{22}y_0$$

or

$$\begin{pmatrix} a_{11} - \lambda & a_{12} \\ a_{21} & a_{22} - \lambda \end{pmatrix} \begin{pmatrix} x_0 \\ y_0 \end{pmatrix} = 0$$



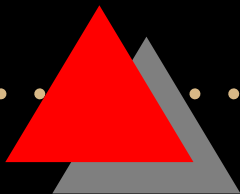


# *Linearization*

From this, we obtain the characteristic equation

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

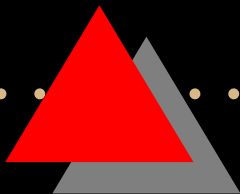
Solving for the two roots of  $\lambda$  will determine the stability of the system.





# *Linearization*

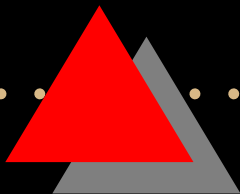
- If both roots of  $\lambda$  are real and negative, the equilibrium is a stable node.
- If both roots of  $\lambda$  are real and positive, the equilibrium is an unstable node.
- If the roots of  $\lambda$  are real and of opposite signs, the equilibrium is a saddle point.





# *Linearization*

- If the roots of  $\lambda$  are complex with negative real parts, the equilibrium is a stable focus.
- If the roots of  $\lambda$  are complex with positive real parts, the equilibrium is an unstable focus.
- If the roots of  $\lambda$  are purely complex, the equilibrium of the linearized system is a center, but the original nonlinear system will have a center or a stable or unstable focus depending upon the exact nature of the nonlinear terms.



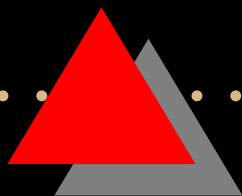




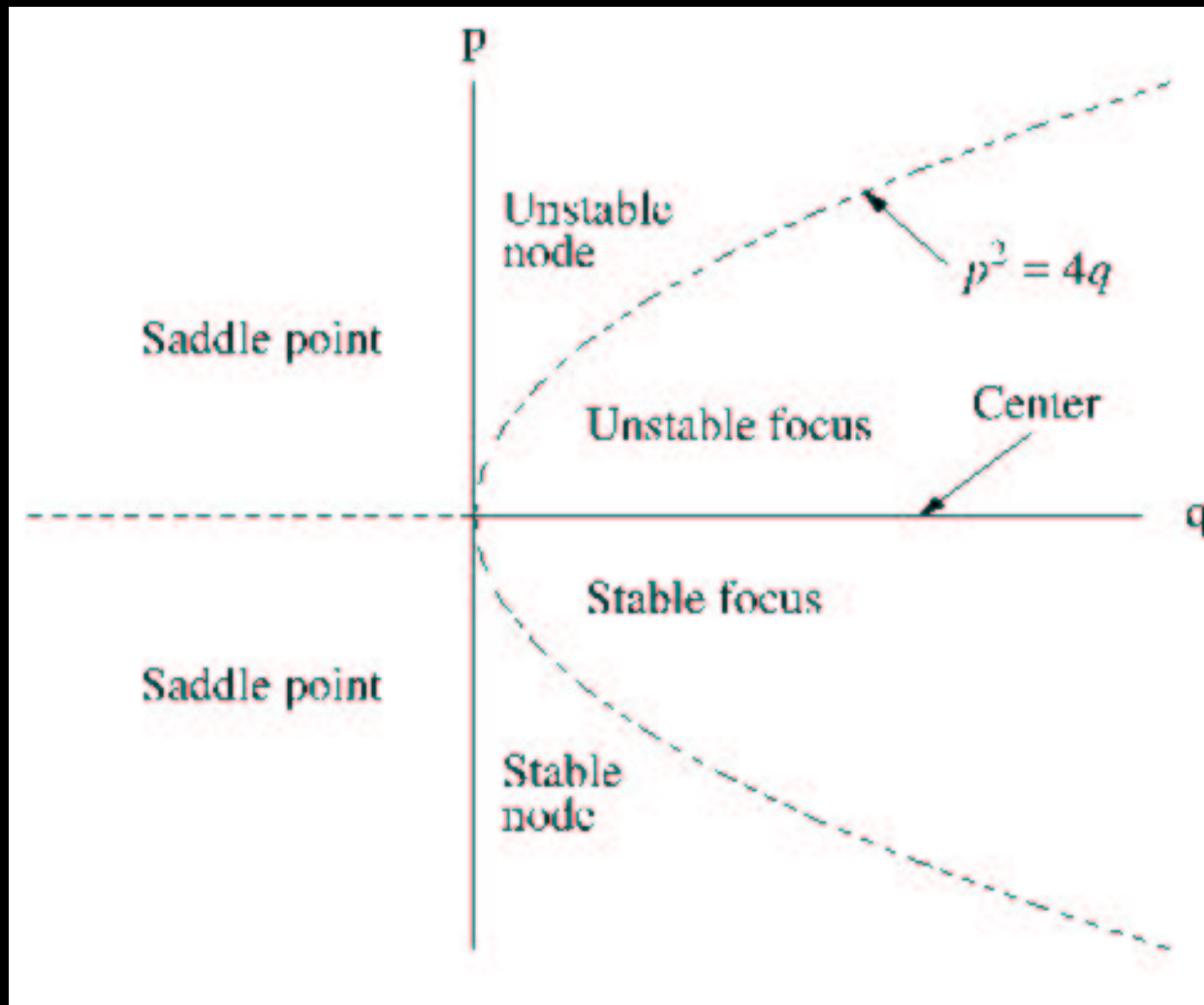
# Routh-Hurwitz conditions

Routh-Hurwitz conditions give the necessary and sufficient conditions for all roots of the characteristic polynomial to have negative real roots thus implying asymptotic stability.

$$\begin{aligned} p = \text{Tr} J &= a_{11} + a_{22} < 0 \\ q = \det J &= a_{11}a_{22} - a_{12}a_{21} > 0 \end{aligned}$$



# Stability





# *Bendixson negative criterion*

Bendixson's negative criterion

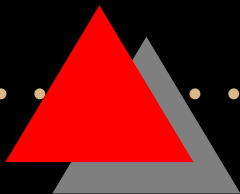
Consider the dynamical system,  $\frac{dx}{dt} = F(x, y), \frac{dy}{dt} = G(x, y)$ , where  $F$  and  $G$  are continuously differentiable functions on some simply connected domain  $D \subset \mathbb{R}^2$ . If  $\nabla \cdot (F, G) = \frac{\partial F}{\partial x} + \frac{\partial G}{\partial y}$  is of one sign in  $D$ , there cannot be a closed orbit contained within  $D$ .



# *Bendixson-Dulac negative criterion*

Additionally, we have the Bendixson's-Dulac's negative criterion.

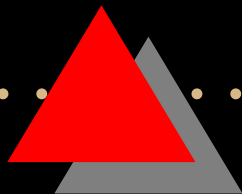
Let  $B$  be a smooth function on  $D \subset \mathbb{R}^2$  (with above assumptions). If  $\nabla \cdot (BF, BG) = \frac{\partial BF}{\partial x} + \frac{\partial BG}{\partial y}$  is of one sign in  $D$ , there cannot be a closed orbit contained within  $D$ .





# *Other theorems*

- The Hopf bifurcation theorem gives conditions necessary for the existence of real periodic solutions of a real system of ordinary differential equations.
- Poincaré-Bendixson theorem can also be used to prove the existence of periodic orbits.



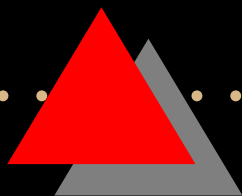


# *Examples from Epidemiology*

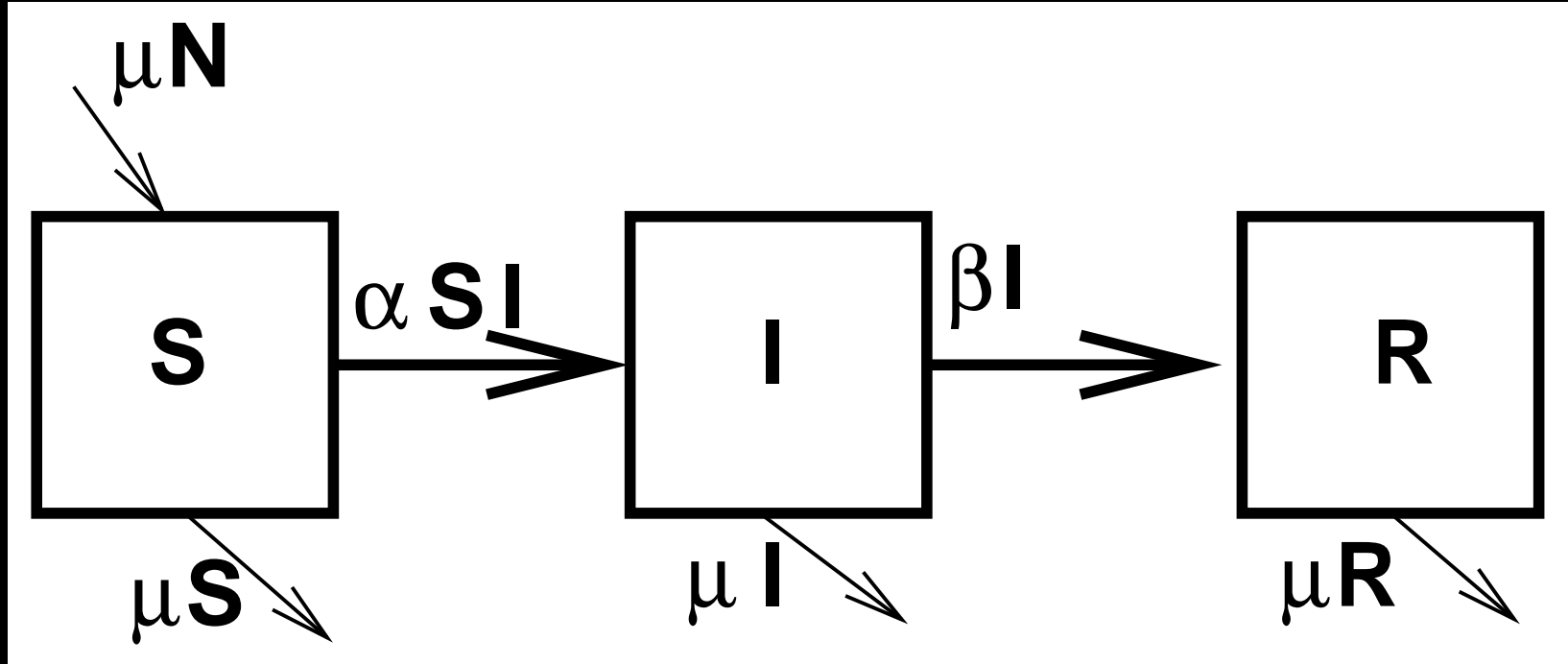
Divide population up into distinct classes

- $S$  = Susceptibles
- $I$  = Infectives
- $R$  = Recovered

Classes used depend on disease dynamics



# *SIR Model*





# *SIR Model - constant population*

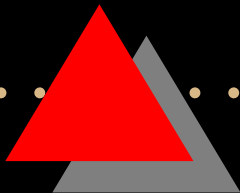
$$\frac{dS}{dt} = \mu(S + I + R) - \alpha SI - \mu S$$

$$\frac{dI}{dt} = \alpha SI - \beta I - \mu I$$

$$\frac{dR}{dt} = \beta I - \mu R$$

$$N = S + I + R$$

$S(0) = S_0, I(0) = I_0, R(0) = 0$ . All parameters are assumed to be positive.



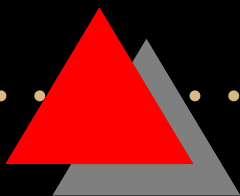




# *Questions for Epidemic Models*

Given all parameters and initial conditions

- Does the infection spread or die out?
- If it does spread, how does it develop with time?
- When will it start to decline?



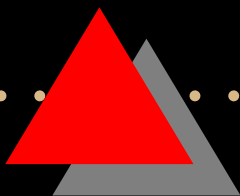


# Equilibria

Note: Since  $N$  is a constant, we can solve for only  $S$  and  $I$ , then if we need  $R$ , we can calculate it easily.

$$\frac{dS}{dt} = \mu N - \alpha SI - \mu S = 0$$

$$\frac{dI}{dt} = \alpha SI - \beta I - \mu I = 0$$



# Equilibria

Gives two equilibria:

$$S^* = N, \quad I^* = 0$$
$$S^* = \frac{\beta + \mu}{\alpha}, \quad I^* = \frac{\mu(\alpha N - \beta - \mu)}{\alpha(\beta + \mu)}$$

Let

$$F(S, I) = \mu N - \alpha SI - \mu S$$

$$G(S, I) = \alpha SI - \beta I - \mu I$$

# Stability

Then

$$\frac{\partial F}{\partial S} = -\alpha I - \mu$$

$$\frac{\partial F}{\partial I} = -\alpha S$$

$$\frac{\partial G}{\partial S} = \alpha I$$

$$\frac{\partial G}{\partial I} = \alpha S - \beta - \mu$$

# Stability

First, let's evaluate the stability of  $S^* = N, I^* = 0$   
The elements of the Jacobian evaluated at this equilibrium are:

$$a_{11} = -\mu$$

$$a_{12} = -\alpha N$$

$$a_{21} = 0$$

$$a_{22} = \alpha N - \beta - \mu$$

# Stability

Applying the Routh-Hurwitz conditions:

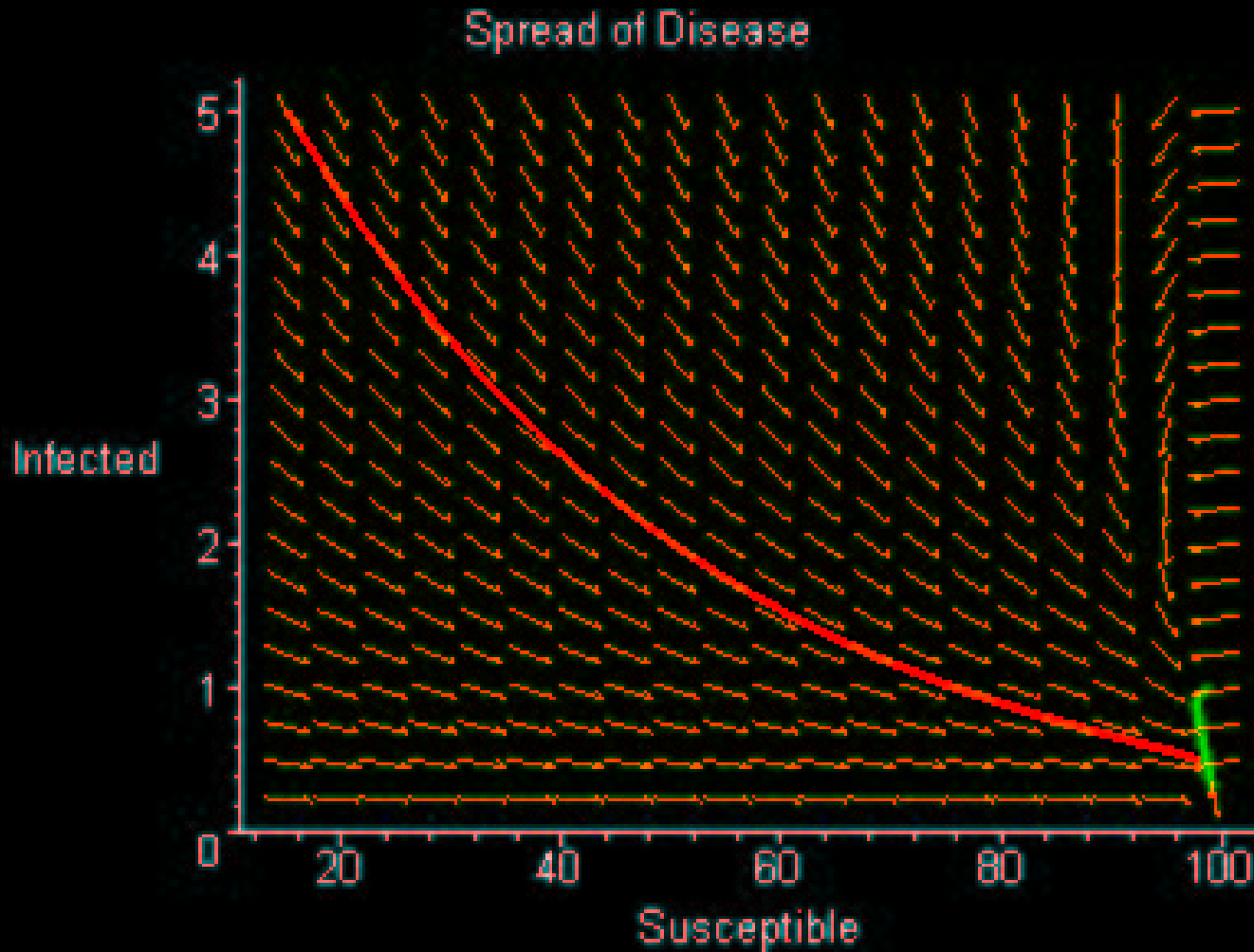
$$\begin{aligned}a_{11} + a_{22} &= -\beta - 2\mu \\ a_{11}a_{22} - a_{12}a_{21} &= \mu(\beta + \mu - \alpha N)\end{aligned}$$

Clearly,  $-\beta - 2\mu < 0$

However,  $\mu(\beta + \mu - \alpha N) > 0$  only if  $\alpha N < \beta + \mu$

Therefore,  $S^* = N, I^* = 0$  is asymptotically stable  
if  $\alpha N < \beta + \mu$

# *Disease dies out*



# Stability

Now, let's evaluate the stability of

$$S^* = \frac{\beta + \mu}{\alpha}, I^* = \frac{\mu(\alpha N - \beta - \mu)}{\alpha(\beta + \mu)}$$

The elements of the Jacobian evaluated at this equilibrium are:

$$a_{11} = -\mu - \frac{\mu(\alpha N - \beta - \mu)}{\beta + \mu}$$

$$a_{12} = -\beta - \mu$$

$$a_{21} = \frac{\mu(\alpha N - \beta - \mu)}{\beta + \mu}$$

$$a_{22} = 0$$



# Stability

Applying the Routh-Hurwitz conditions:

$$a_{11} + a_{22} = -\mu - \frac{\mu(\alpha N - \beta - \mu)}{\beta + \mu}$$

$$a_{11}a_{22} - a_{12}a_{21} = (\beta + \mu) \left( \frac{\mu(\alpha N - \beta - \mu)}{\beta + \mu} \right)$$

# Stability

Both

$$-\mu - \frac{\mu(\alpha N - \beta - \mu)}{\beta + \mu} < 0$$

$$(\beta + \mu) \left( \frac{\mu(\alpha N - \beta - \mu)}{\beta + \mu} \right) < 0$$

are true if  $\alpha N > \beta + \mu$

# Stability

Therefore,

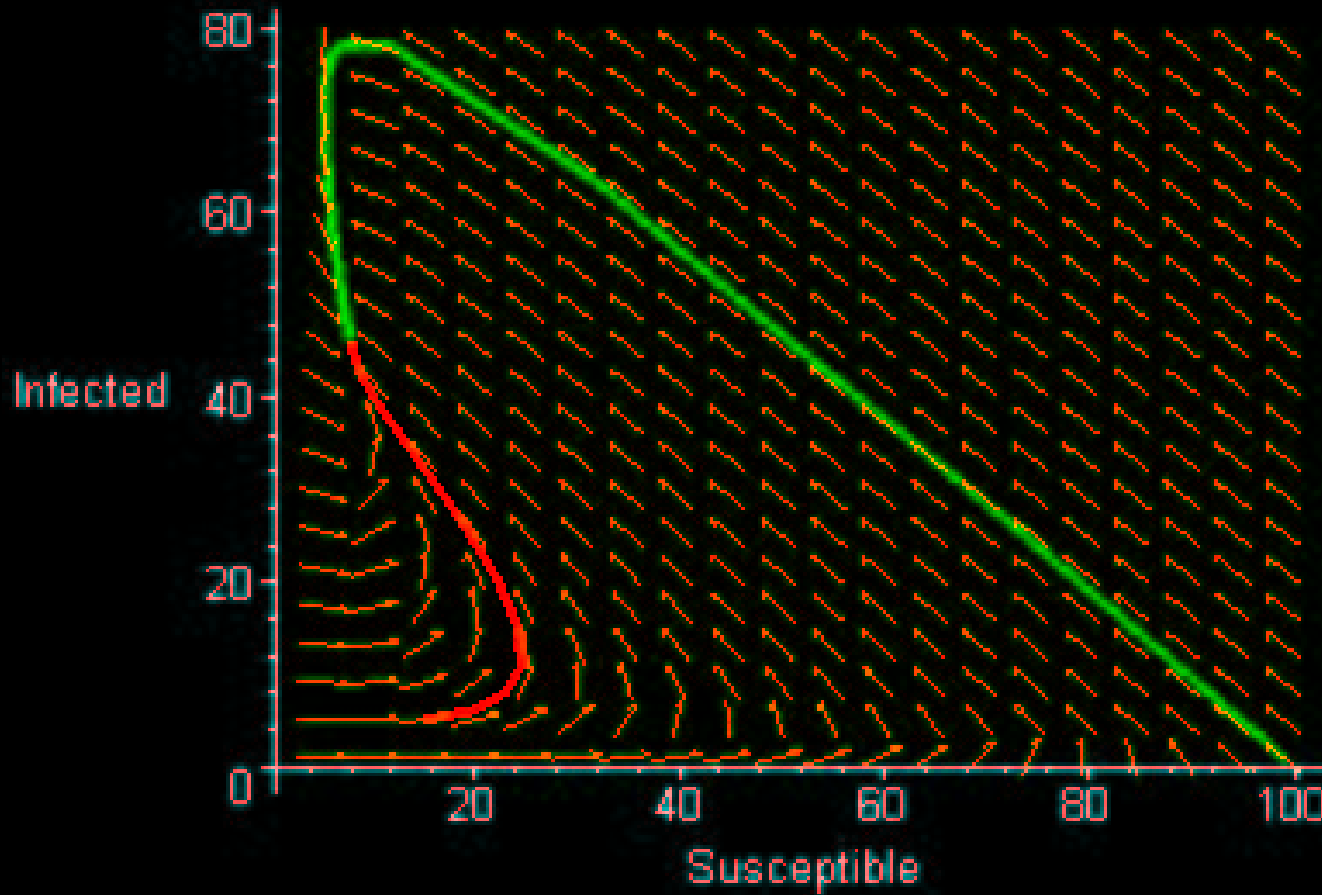
$$S^* = \frac{\beta + \mu}{\alpha}, \quad I^* = \frac{\mu(\alpha N - \beta - \mu)}{\alpha(\beta + \mu)}$$

is asymptotically stable if  $\alpha N < \beta + \mu$

But what about limit cycles?

# Epidemic

Spread of Disease





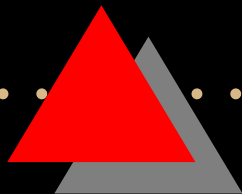
# *Bendixson's Negative Criteria*

Recall, we need

$$\nabla \cdot (F, G) = \frac{\partial F}{\partial S} + \frac{\partial G}{\partial I}$$

to be of one sign in our region of interest,  $D$ .

Define  $D$  to be all positive values in  $\mathbb{R}^2$ .



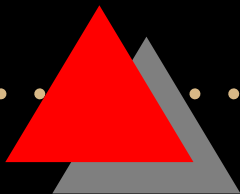


# *Bendixson's Negative Criteria*

$$\frac{\partial F}{\partial S} + \frac{\partial G}{\partial I} = -\alpha I - \mu + \alpha S - \beta - \mu$$

So this will be of one sign, negative, if  $\alpha N < \beta + \mu$  since  $S \leq N$ .

Therefore there are no limit cycles in  $D$ .





# $R_0$ Basic reproduction rate

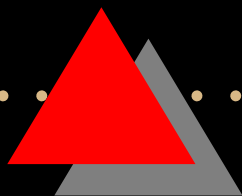
- $R_0$  is defined to be the number of secondary infections produced by one primary infection in a wholly susceptible population.
- So if  $R_0 > 1$ , then the disease will spread.
- For SIR model,  $R_0$  is calculated by linearizing the equation for  $\frac{dI}{dt}$  about  $I = 0$ , which we have already done.
- So the criteria for determining if the epidemic will spread is,  $R_0 \equiv \frac{\alpha N}{\beta + \mu}$ .



# *Conclusions*

There are many other applications of differential equation models in biology. Once a basic set of equations has been developed, there are a number of standard techniques used to analyze the stability of the equations.

We will take time in the lab to explore these and other equations.







# *Acknowledgements*

- Murray, J. D., *Mathematical Biology*, Springer-Verlag, 1989.
- Kot, Mark, *Elements of Mathematical Ecology*, Cambridge University Press, 2001.
- Arrowsmith, D.K. and C.M. Place, *Ordinary Differential Equations*, Chapman and Hall, 1982.

All slides created in  $\text{L}^{\text{A}}\text{T}_{\text{E}}\text{X}$  using the Prosper class.

