

Introduction to system models

Definition

- A system
 - a set of interacting components grouped together in order to study some part of the real world
- A mathematical model
 - A simplified mathematical representation of the relationship between variables
- A (mathematical) system model
 - A representation of a system in equations
 - Usually dynamic equations to describe the dynamics of the system



An example

predator prey interaction

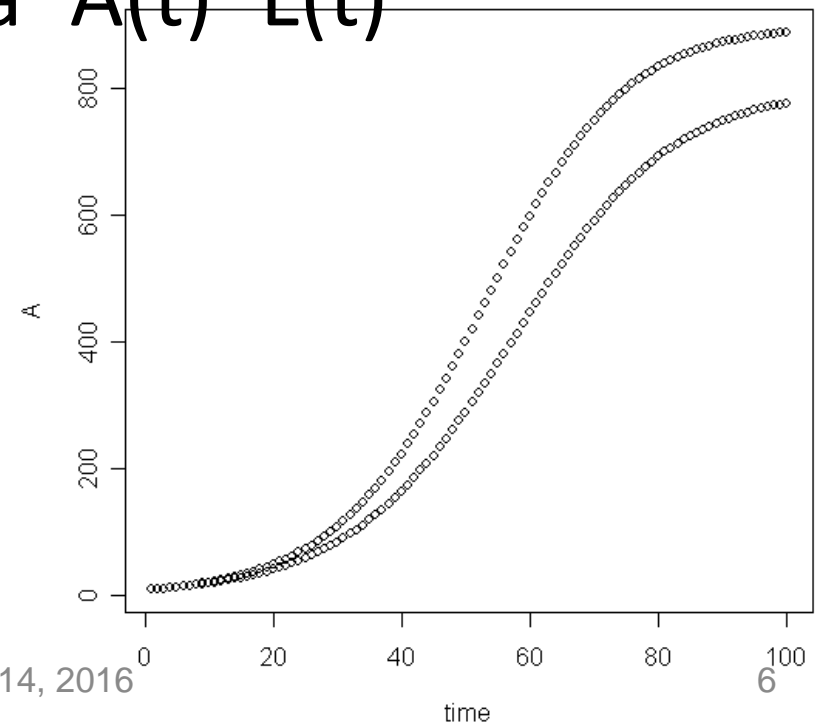


Predator prey model

- $A(t)$ =number of individuals of prey at time t
($A(t)$ =aphids at time t)
- $L(t)$ =number of individuals of predator at time t
($L(t)$ =ladybugs at time t)

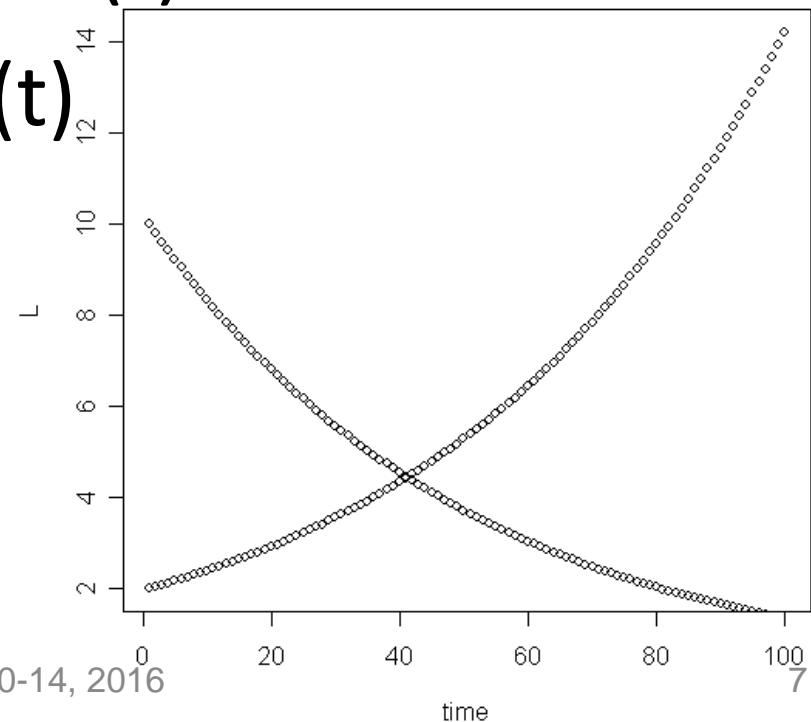
Model for prey (for given number of predators)

- Prey have logistic growth, plus mortality due to predation
- Predation proportional to $A(t)$ and to $L(t)$
- $dA/dt = r_A * A(t) * (1 - A(t)/K) - a * A(t) * L(t)$



Model for predator (for given number of prey)

- Increase depends on food supply, proportional to $A(t)$ and to $L(t)$
- Mortality is proportional to $L(t)$
- $dL(t)/dt = b * A(t) * L(t) - r_L * L(t)$



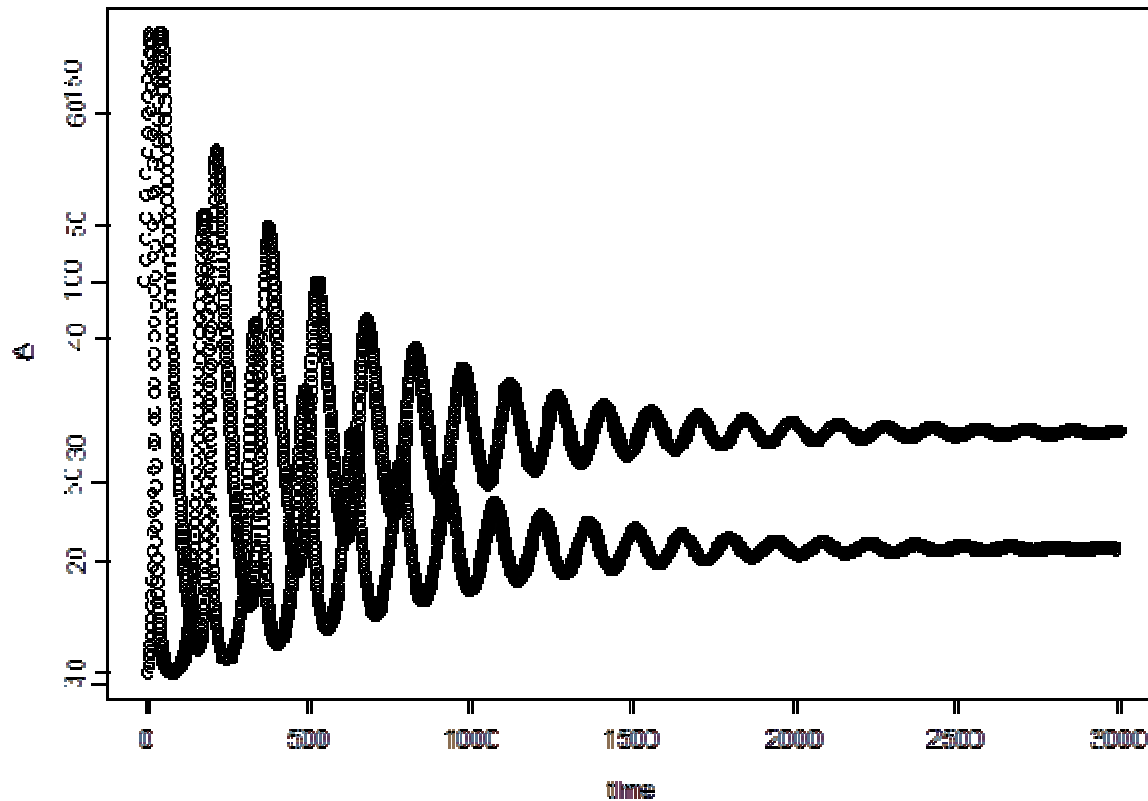
Predator and prey together

- Previously
 - Prey dynamics for fixed number of predators
 - Predator dynamics for fixed number of prey
- Now include interaction.
 - Solve equations simultaneously
 - This is a (dynamic) system model (dynamic equations for two components, A and L, that interact)

$$dA(t)/dt = r_A * A(t) * (1 - A(t)/K) - a * A(t) * L(t)$$

$$dL(t)/dt = b * A(t) * L(t) - r_L * L(t)$$

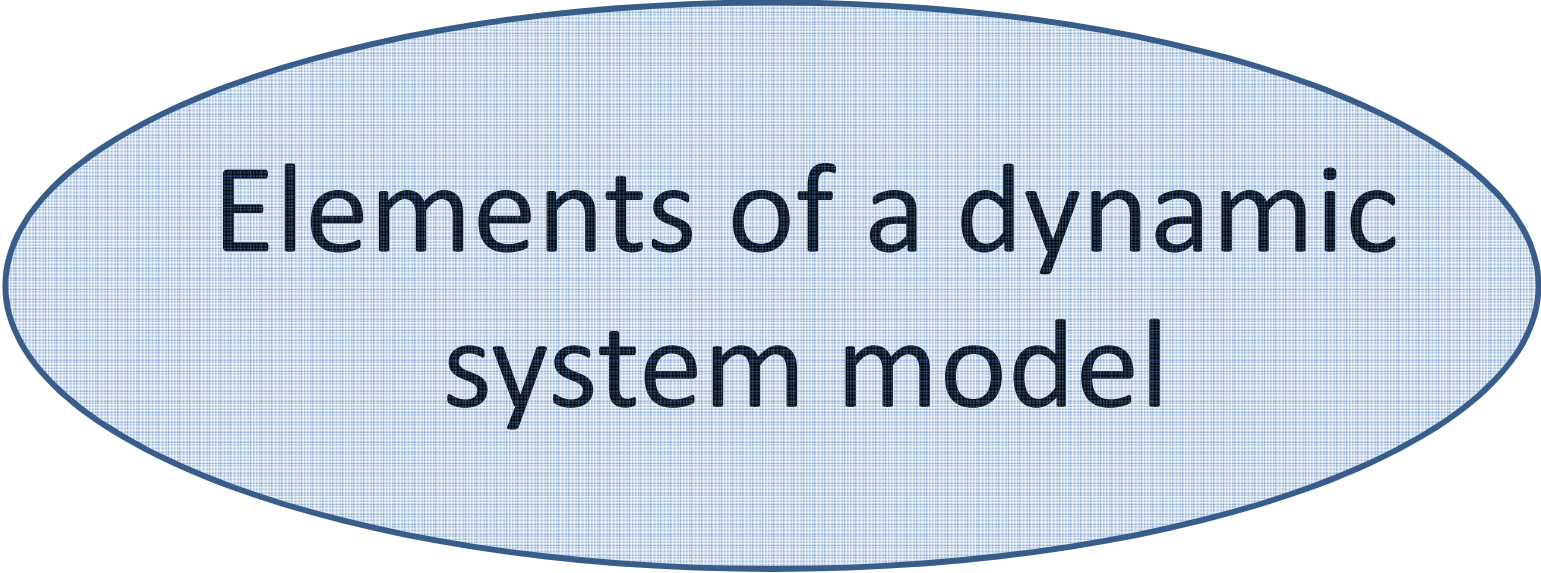
Predator and prey together



Out-of-phase oscillations.

Importance of systems approach

- Study of components is necessary but not sufficient to understand system behavior
- The system can have behavior that you don't see by studying individual components.



Elements of a dynamic system model

State variables

- These are the variables whose dynamics are calculated
 - In our case $A(t)$, $L(t)$
- Easily recognized, they appear on left of equations, as $d(\text{state variable})/dt$

$$dA(t)/dt = r_A * A(t) * (1 - A(t)/K) - a * A(t) * L(t)$$

$$dL(t)/dt = b * A(t) * L(t) - r_L * L(t)$$

- The choice of state variables is very important
- That defines what is included in the system
 - We only calculate the dynamics of aphids and ladybugs, not other populations, or plants etc. The choice of state variables is very important

Explanatory variables

- These are measurable variables that affect dynamics of system, but aren't affected by system.
 - Appear only on right of equations, plus initial values
 - Here, only the initial values $A(t=0)$ and $L(t=0)$
 - No time dependent explanatory variables
 - Could have temperature for example

$$dA(t)/dt = r_A * A(t) * (1 - A(t)/K) - a * A(t) * L(t)$$

$$dL(t)/dt = b * A(t) * L(t) - r_L * L(t)$$

- The choice of explanatory variables is very important
 1. Because to simulate with the model, we need to know those values
 - This determines whether model can be used in practice. If we don't know $A(t=0)$ and $L(t=0)$, can't simulate dynamics of system
 2. Because it sets an upper limit on how good the model can be
 - Here, no weather variables, no plant variables, no variables for other populations. The model can't describe variability due to ignored explanatory variables.

Parameters

- Not calculated by the model and not measured for each field.
- Here there are five: r_A , K , a , b , r_L

$$dA(t)/dt = r_A * A(t) * (1 - A(t)/K) - a * A(t) * L(t)$$

$$dL(t)/dt = b * A(t) * L(t) - r_L * L(t)$$

- Where do the values of the parameters come from?
 - Not calculated by the model, not measured
 - They come from past studies of similar systems.
 - We can use past values because we assume parameters don't change. Same values for all fields (or some subset of fields).

- The estimation of parameter values is very important
 - Different values can give very different predictions.
 - This is a major difficulty in modeling (system or other)

Output variables

- Often, not interested in full dynamics, but in some specific output.
 - e.g. yield of a crop, integrated damage of a disease like $\int A(t)dt$
 - In that case, we can write the model as $\hat{y}=f(X;\theta)$
 - \hat{y} is simulated output. It is calculated as function of explanatory variables X and parameters θ .

- It is important to define the outputs of interest.
 - Quality of model can be very different for different outputs.

What is special about system models?

- If we look at one output, a system model is just a regression model
 - Relates output to explanatory variables
 - Standard regression equation is $Y=f(X;\theta)+\varepsilon$
 - Observed Y is simulated value plus error
 - So what's special?

Specificity of system models

- We have two complementary ways of studying dynamic system models
 - We can study individual processes
 - We can study the overall system
 - This is a major difference with other models

1. We can study the individual processes
 - rate of increase of aphids in absence of ladybugs
 - rate of predation versus density of aphids
 - natural mortality of ladybugs
 - effect of predation rate on rate of increase of ladybugs
2. We can study full system
 - Aphids and ladybugs interacting.

- Those involve quite different experiments
- A major challenge in system modeling is combining these two sources of information.
 - In particular, we have two ways of getting the parameter values
 - We can estimate them either by studying the individual processes or by studying the overall system

