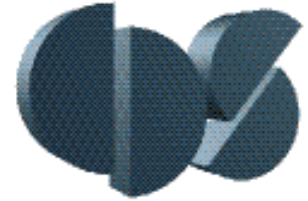




# CDS 101: Lecture 2.1

## System Modeling



**Richard M. Murray**

**4 October 2004**

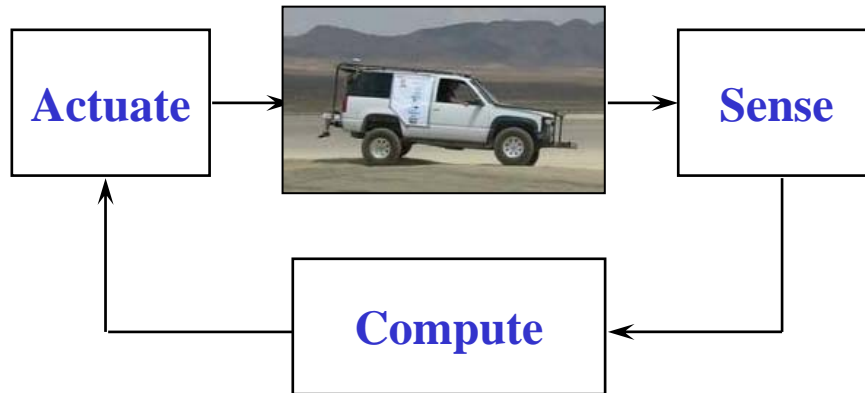
### **Goals:**

- Define what a model is and its use in answering questions about a system
- Introduce the concepts of state, dynamics, inputs and outputs
- Provide examples of common modeling techniques: differential equations, difference equations, finite state automata

### **Reading:**

- Åström and Murray, *Analysis and Design of Feedback Systems*, Ch 2
- Advanced: Lewis, *A Mathematical Approach to Classical Control*, Ch 1

# Review from last week



**Control =**

Sensing + Computation +  
Actuation

**Feedback Principles**

- Robustness to Uncertainty
- Design of Dynamics

**Many examples of feedback and control in natural & engineered systems:**

A collage of four images illustrating feedback and control in different systems:

- Mechanical:** A diagram of a mechanical governor, showing a flyball mechanism with a 'Control valve' and 'Steam' input, and a 'Drive belt' at the bottom.
- BIO:** A biological diagram showing a cross-section of a cell or tissue with various components labeled 'Ch1', 'Ch2', 'Ch3', and 'Ch4'.
- ESE:** A diagram of a power system showing a 'Generator' and 'Load' connected to a 'Transmission Line'.
- CS:** A network diagram showing a complex web of connections between various nodes, representing a control system in computer science.
- Electrical:** A detailed circuit diagram with a 25V source, resistors (150K, 10K, 47K, 4.7K, 33K, 4.7K, 4.7K, 4.7K, 100Ω), capacitors (6.8pF, 50pF, 60pF, 10pF, 5pF), and transistors (Q1, Q2).
- Natural:** A photograph of a fly, illustrating natural control systems.

# Model-Based Analysis of Feedback Systems

## Analysis and design based on *models*

- A model provides a *prediction* of how the system will behave
- Feedback can give counter-intuitive behavior; models help sort out what is going on
- For control design, models don't have to be exact: *feedback* provides robustness

## Control-oriented models: *inputs* and *outputs*

## The model you use depends on the questions you want to answer

- A single system may have many models
- Time and spatial scale must be chosen to suit the questions you want to answer
- Formulate questions *before* building a model

## Weather Forecasting



- Question 1: how much will it rain tomorrow?
- Question 2: will it rain in the next 5-10 days?
- Question 3: will we have a drought next summer?

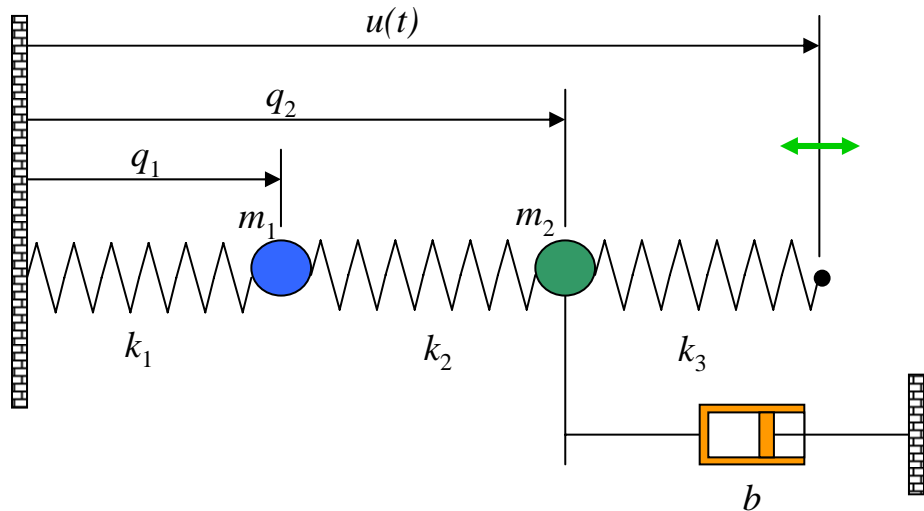
**Different questions  $\Rightarrow$  different models**

**Slide 4**

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**RMM16** Use a different model, more related to feedback and control  
Richard Murray, 10/4/2003

# Example #1: Spring Mass System



## Applications

- Flexible structures (many apps)
- Suspension systems (eg, “Bob”)
- Molecular and quantum dynamics

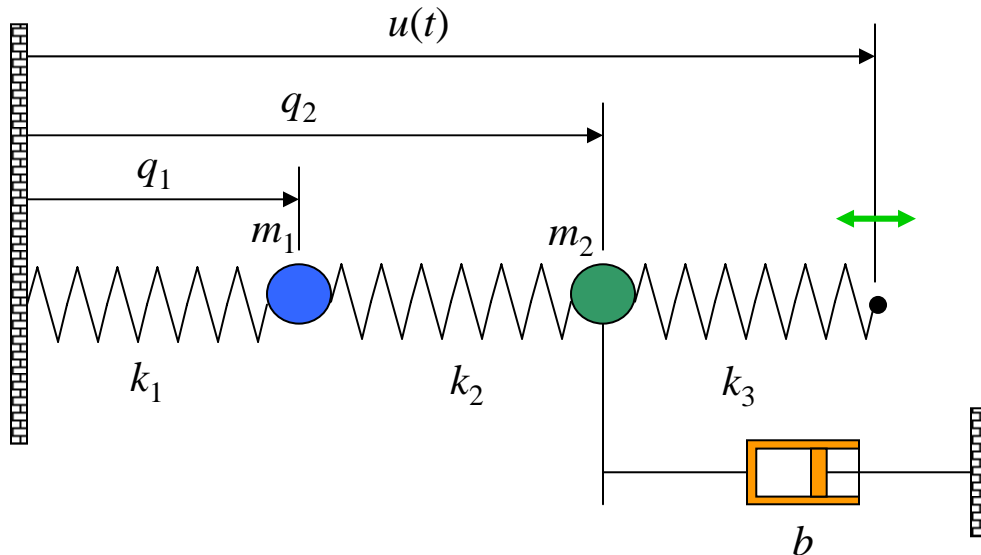
## Questions we want to answer

- How much do masses move as a function of the forcing frequency?
- What happens if I change the values of the masses?
- Will Bob fly into the air if I take that hill at 25 mph?

## Modeling assumptions

- Mass, spring, and damper constants are fixed and known
- Springs satisfy Hooke’s law
- Damper is (linear) viscous force, proportional to velocity

# Modeling a Spring Mass System



## Model: rigid body physics (Ph 1)

- Sum of forces = mass \* acceleration
- Hooke's law:  $F = k(x - x_{\text{rest}})$
- Viscous friction:  $F = b v$

$$\begin{aligned} m_1 \ddot{q}_1 &= k_2(q_2 - q_1) - k_1 q_1 \\ m_2 \ddot{q}_2 &= k_3(u - q_2) - k_2(q_2 - q_1) - b \dot{q}_2 \end{aligned}$$

## Converting models to state space form

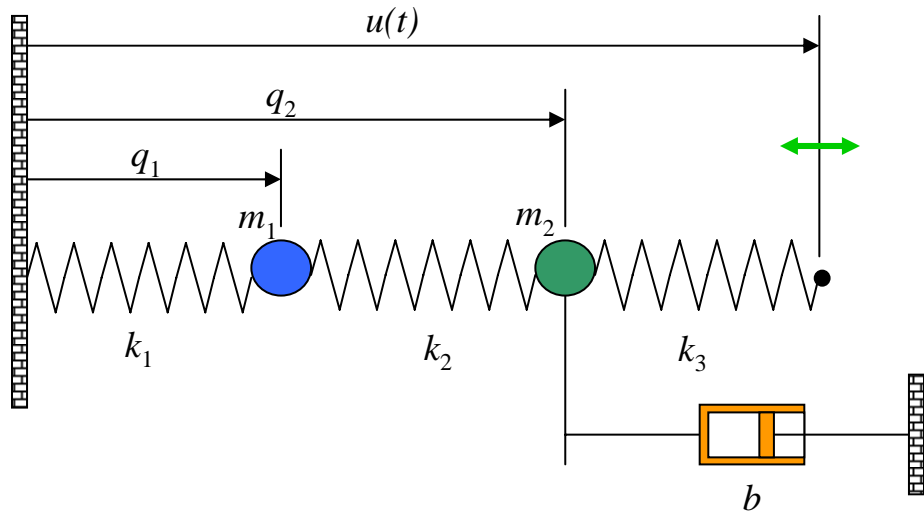
- Construct a *vector* of the variables that are required to specify the evolution of the system
- Write dynamics as a *system* of first order differential equations:

$$\begin{aligned} \frac{dx}{dt} &= f(x, u) \quad x \in \mathbb{R}^n, u \in \mathbb{R}^p \\ y &= h(x) \quad y \in \mathbb{R}^q \end{aligned}$$

$$\frac{d}{dt} \begin{bmatrix} q_1 \\ q_2 \\ \dot{q}_1 \\ \dot{q}_2 \end{bmatrix} = \begin{bmatrix} \dot{q}_1 \\ \dot{q}_2 \\ \frac{k_2}{m}(q_2 - q_1) - \frac{k_1}{m}q_1 \\ \frac{k_3}{m}(u - q_2) - \frac{k_2}{m}(q_2 - q_1) - \frac{b}{m}\dot{q}_2 \end{bmatrix}$$

$$y = \begin{bmatrix} q_1 \\ q_2 \end{bmatrix} \quad \text{“State space form”}$$

# Frequency Response for a Mass Spring System



## Steady state frequency response

- Force the system with a sinusoid
- Plot the “steady state” response, after transients have died out
- Plot relative magnitude and phase of output versus input (more later)

## Matlab simulation (see handout)

**function** dydt = f(t, y, ...)

u = 0.00315\*cos(omega\*t);

dydt = [

  y(3);

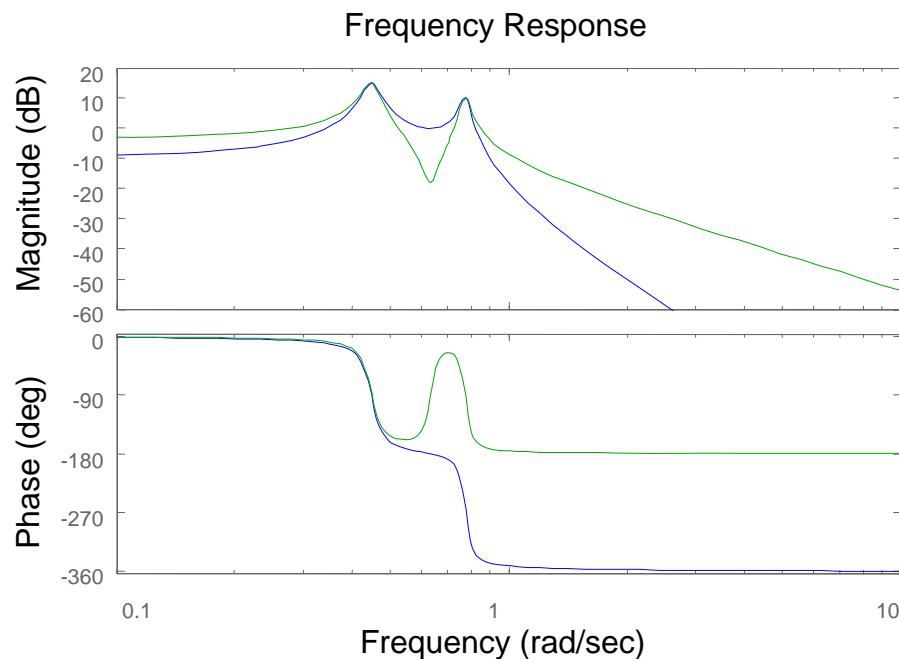
  y(4);

  -(k1+k2)/m1\*y(1) + k2/m1\*y(2);

  k2/m2\*y(1) - (k2+k3)/m2\*y(2)

  - b/m2\*y(4) + k3/m2\*u ];

t,y] = **ode45**(dydt,tspan,y0,[], k1, k2, k3, m1, m2, b, omega);



## Modeling Terminology

### State captures effects of the past

- independent physical quantities that determines future evolution (absent external excitation)

### Inputs describe external excitation

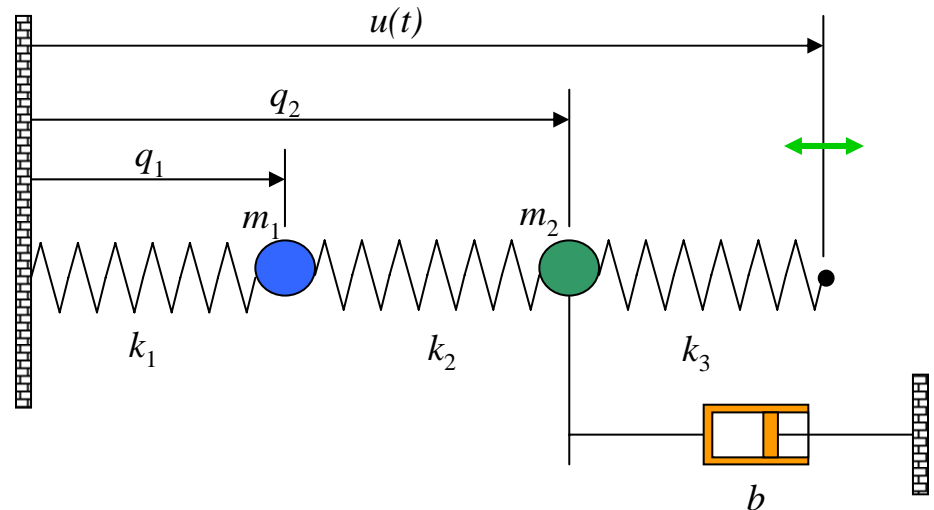
- Inputs are *extrinsic* to the system dynamics (externally specified)

### Dynamics describes state evolution

- update rule for system state
- function of current state and any external inputs

### Outputs describe measured quantities

- Outputs are function of state and inputs  $\Rightarrow$  not independent variables
- Outputs are often *subset* of state



### Example: spring mass system

- State: position and velocities of each mass:  $q_1, q_2, \dot{q}_1, \dot{q}_2$
- Input: position of spring at right end of chain:  $u(t)$
- Dynamics: basic mechanics
- Output: measured positions of the masses:  $q_1, q_2$



**Slide 8**

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**RMM14** Put an example back in (cut out of previous)  
Richard Murray, 10/4/2003

# Modeling Properties

## Choice of state is not unique

- There may be *many* choices of variables that can act as the state
- Trivial example: different choices of units (scaling factor)
- Less trivial example: sums and differences of the mass positions (HW 2.4)

## Choice of inputs and outputs depends on point of view

- Inputs: what factors are *external* to the model that you are building
  - Inputs in one model might be outputs of another model (eg, the output of a cruise controller provides the input to the vehicle model)
- Outputs: what physical variables (often states) can you *measure*
  - Choice of outputs depends on what you can sense and what parts of the component model interact with other component models

## Can also have different *types* of models

- Ordinary differential equations for rigid body mechanics
- Finite state machines for manufacturing, Internet, information flow
- Partial differential equations for fluid flow, solid mechanics, etc

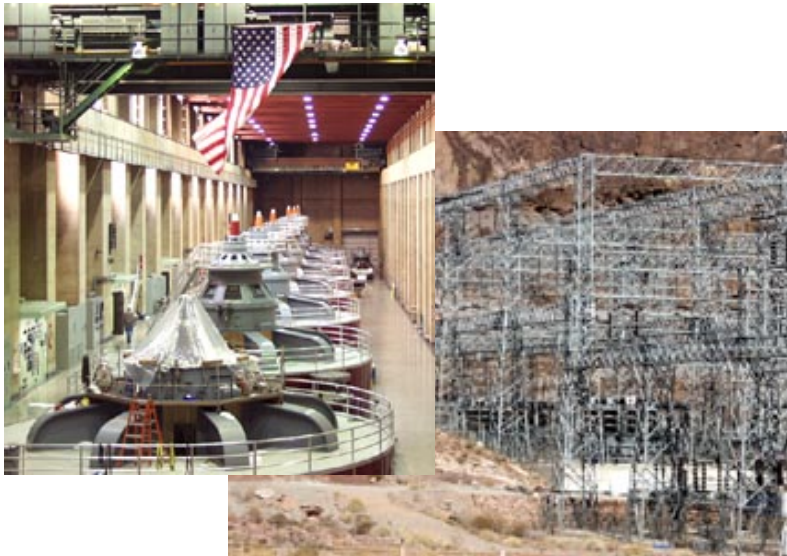
# Differential Equations

## Differential equations model continuous evolution of state variables

- Describe the rate of change of the state variables
- Both state and time are continuous variables

$$\frac{dx}{dt} = f(x, u)$$
$$y = h(x)$$

## Example: electrical power grid



## Swing equations

$$\ddot{\delta}_1 + D_1 \dot{\delta}_1 = \omega_0 (P_1 - B \sin(\delta_1 - \delta_2) + G \cos(\delta_1 - \delta_2))$$

$$\ddot{\delta}_2 + D_2 \dot{\delta}_2 = \omega_0 (P_2 - B \sin(\delta_1 - \delta_2) + G \cos(\delta_1 - \delta_2))$$

- Describe how generator rotor angles ( $\delta_i$ ) interact through the transmission line ( $G, B$ )
- Stability of these equations determines how loads on the grid are accommodated

**State:** rotor angles, velocities ( $\delta_i, \dot{\delta}_i$ )

**Inputs:** power loading on the grid ( $P_i$ )

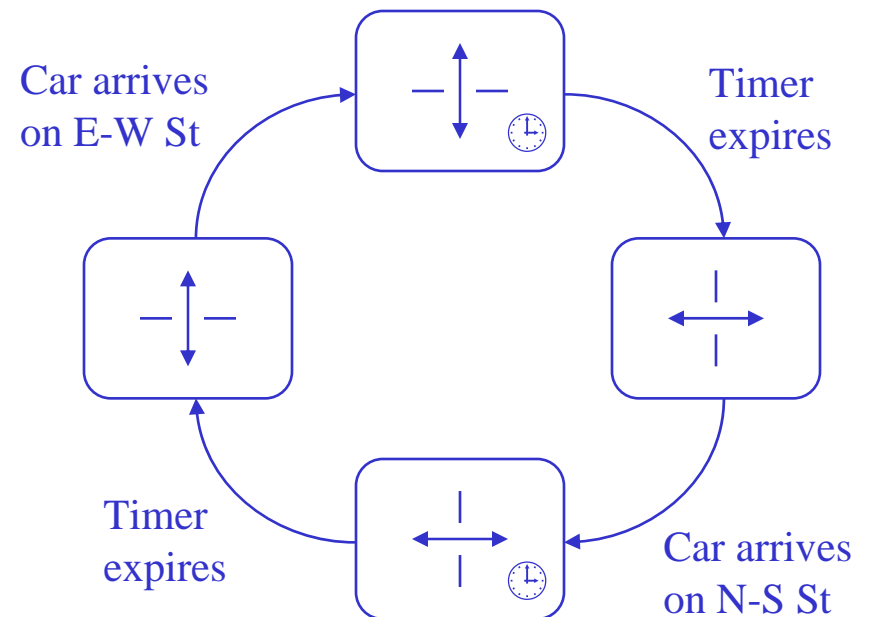
**Outputs:** voltage levels and frequency (based on rotor speed)

# Finite State Machines

**Finite state machines model discrete transitions between finite # of states**

- Represent each configuration of system as a state
- Model transition between states using a graph
- Inputs force transition between states

## Example: Traffic light logic



**State:** current pattern of lights that are on + internal timers

**Inputs:** presence of car at intersections

**Outputs:** current pattern of lights that are on

# Difference Equations

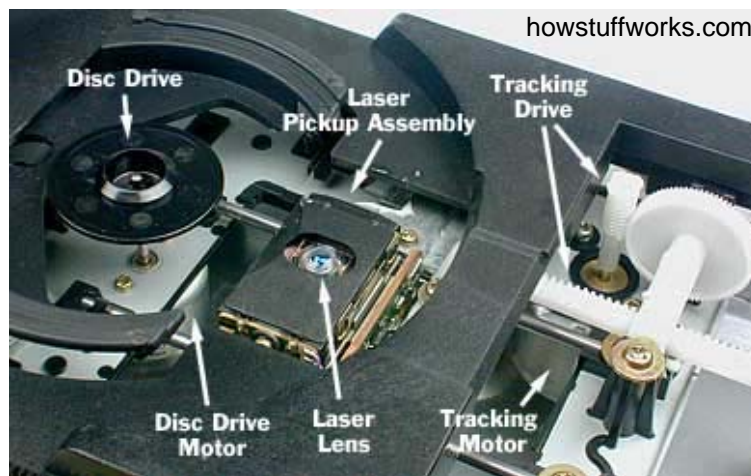
**Difference eqs model discrete transitions between continuous variables**

- “Discrete time” description (clocked transitions)
- New state is function of current state + inputs
- State is represented as a *continuous* variable

$$x_{k+1} = f(x_k, u_k)$$

$$y_{k+1} = h(x_{k+1})$$

**Example: CD read/write head *controller* (implemented on DSP)**



**Controller operation (every 1/44,100 sec)**

- Get analog signal from read head
- Determine the data (1/0) plus estimate the location of the track center
- Update estimate of “wobble”
- Compute where to position disk head for next read (limited by motor torque)

**Performance specification**

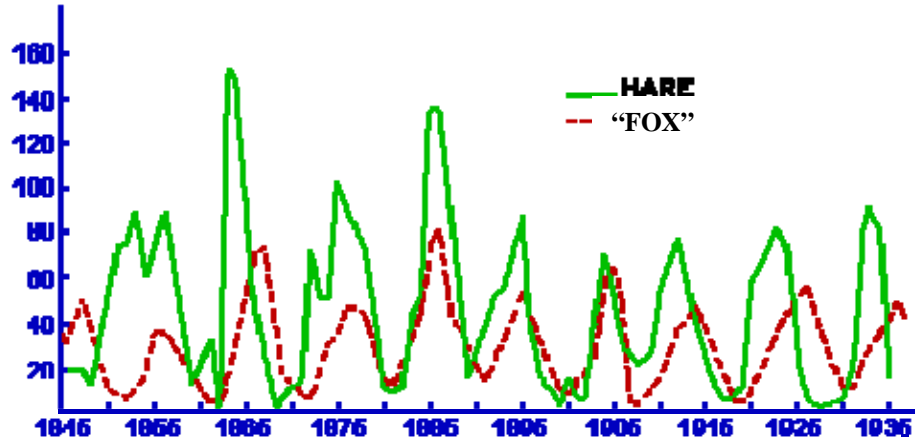
- Keep disk head on track center
- Reject disturbances due to disk shape, shaking and bumps, etc

**State:** estimated center, wobble

**Inputs:** read head signal

**Outputs:** commanded motion

## Example #2: Predator Prey



<http://www.math.duke.edu/education/ccp/materials/diffeq/predprey/contents.html>

### Questions we want to answer

- Given the current population of rabbits and foxes, what will it be next year?
- If we hunt down lots of foxes in a given year, what will the effect on the rabbit and fox population be?
- How do long term changes in the amount of rabbit food available affect the populations?

### Modeling assumptions

- The predator species is totally dependent on the prey species as its only food supply.
- The prey species has an external food supply and no threat to its growth other than the specific predator.

**Slide 13**

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**RMM7**

Tie to Hideo's Friday lecture, if possible

Richard Murray, 9/4/2002

## Example #2: Predator Prey (2/2)

### Discrete Lotka-Volterra model

- State
  - $R_k$  # of rabbits in period  $k$
  - $F_k$  # of foxes in period  $k$
- Inputs (optional)
  - $u_k$  amount of rabbit food
- Outputs: # of rabbits and foxes
- Dynamics: Lotka-Volterra eqs

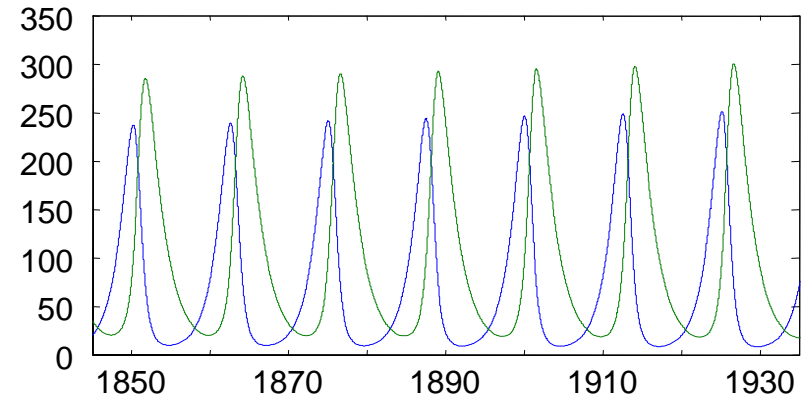
$$R_{k+1} = R_k + b_r(u)R_k - aF_kR_k$$

$$F_{k+1} = F_k - d_fF_k + aF_kR_k$$

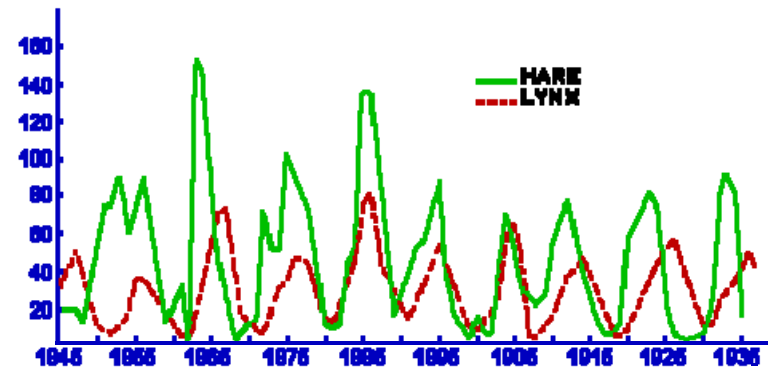
- Parameters/functions
  - $b_r(u)$  rabbit birth rate (per year)  
(depends on food supply)
  - $d_f$  fox death rate (per year)
  - $a$  interaction term

### Matlab simulation (see handout)

- Discrete time model, “simulated” through repeated addition



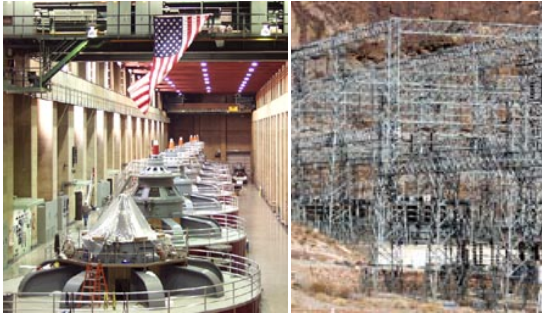
### Comparison with data





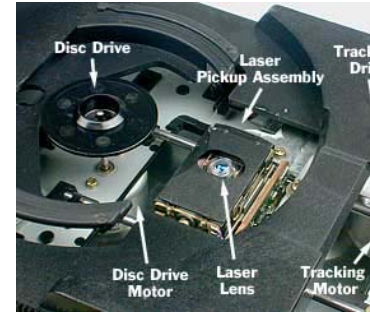
# Summary: System Modeling

Model = state, inputs, outputs, dynamics



$$\frac{dx}{dt} = f(x, u)$$

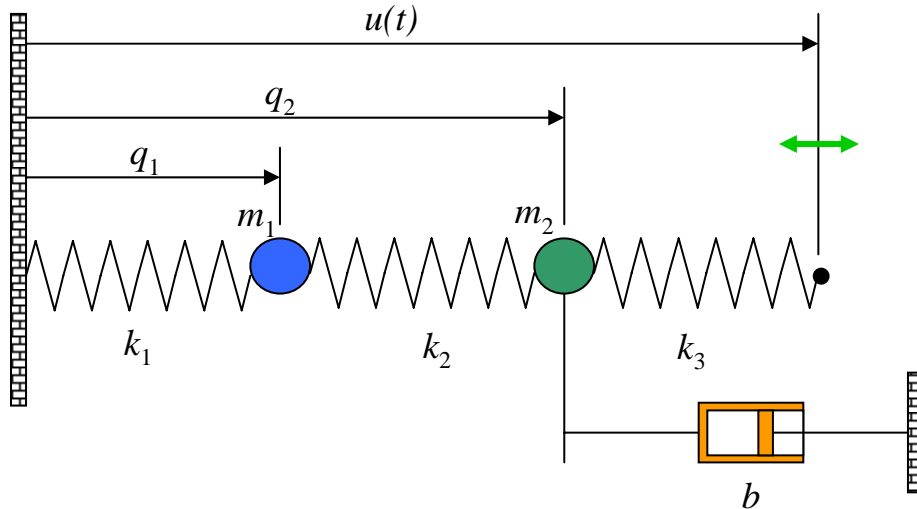
$$y = h(x)$$



$$x_{k+1} = f(x_k, u_k)$$

$$y_{k+1} = h(x_{k+1})$$

**Principle:** Choice of model depends on the questions you want to answer



```
function dydt = f(t, y, k1, k2,
k3, m1, m2, b, omega)
u = 0.00315*cos(omega*t);
dydt = [
    y(3);
    y(4);
    -(k1+k2)/m1*y(1) +
        k2/m1*y(2);
    k2/m2*y(1) - (k2+k3)/m2*y(2)
    - b/m2*y(4) + k3/m2*u ];
```