

# Introduction to Simulation - Lecture 12

## **Methods for Ordinary Differential Equations**

Jacob White

Thanks to Deepak Ramaswamy, Jaime Peraire, Michal  
Rewienski, and Karen Veroy

# Outline

Initial Value problem examples

Signal propagation (circuits with capacitors).

Space frame dynamics (struts and masses).

Chemical reaction dynamics.

Investigate the simple finite-difference methods

Forward-Euler, Backward-Euler, Trap Rule.

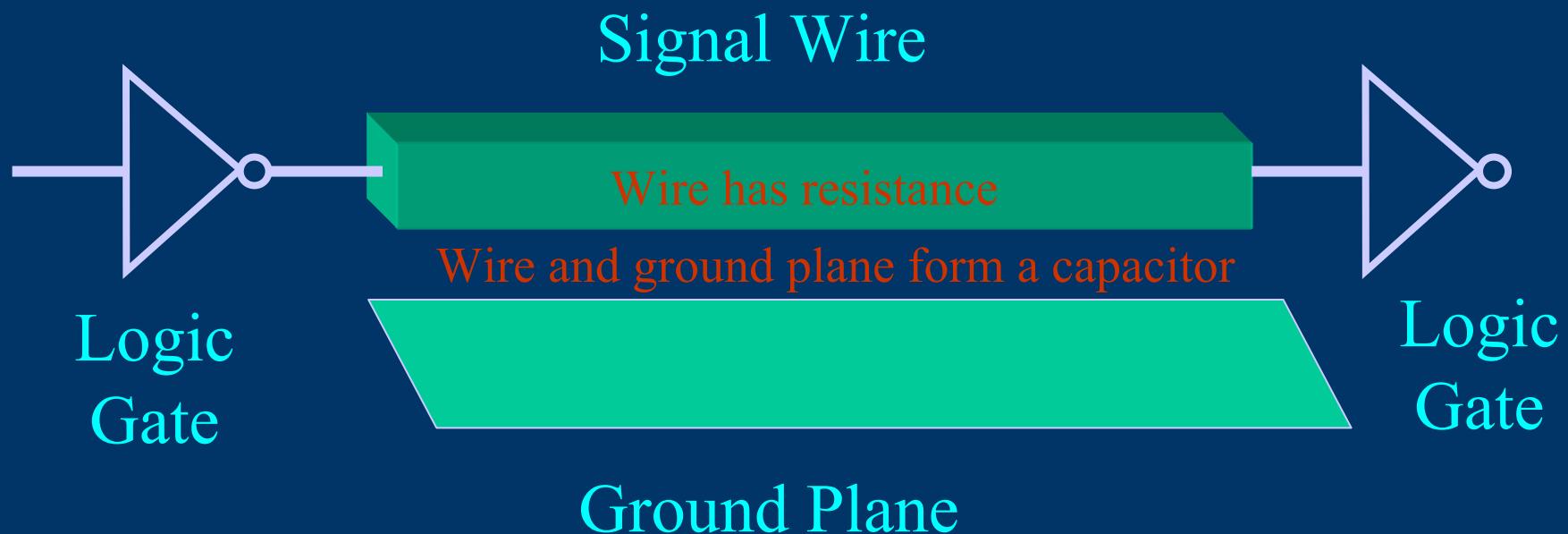
Look at the approximations and algorithms

Examine properties experimentally.

Analyze Convergence for Forward-Euler

# Application Problems

## Signal Transmission in an Integrated Circuit

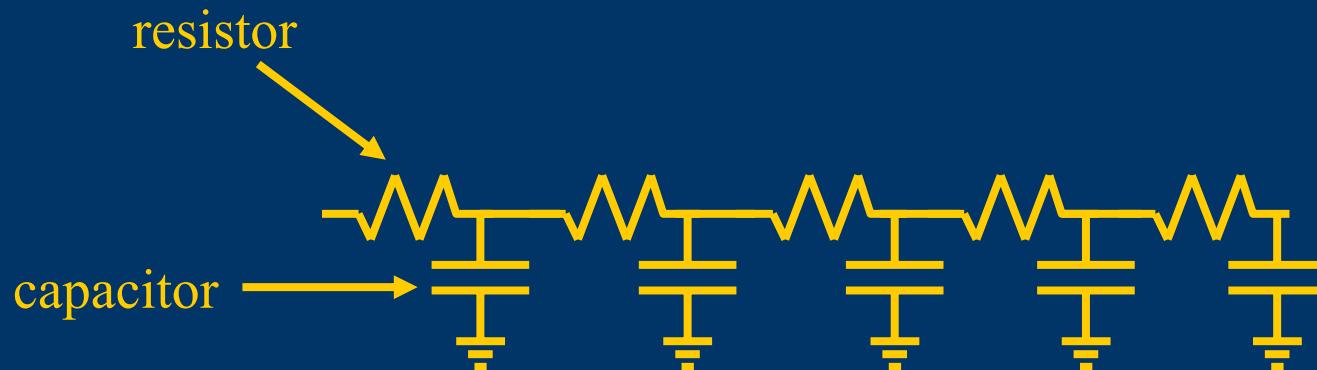


- Metal Wires carry signals from gate to gate.
- How long is the signal delayed?

# Application Problems

## Signal Transmission in an Integrated Circuit

### Circuit Model

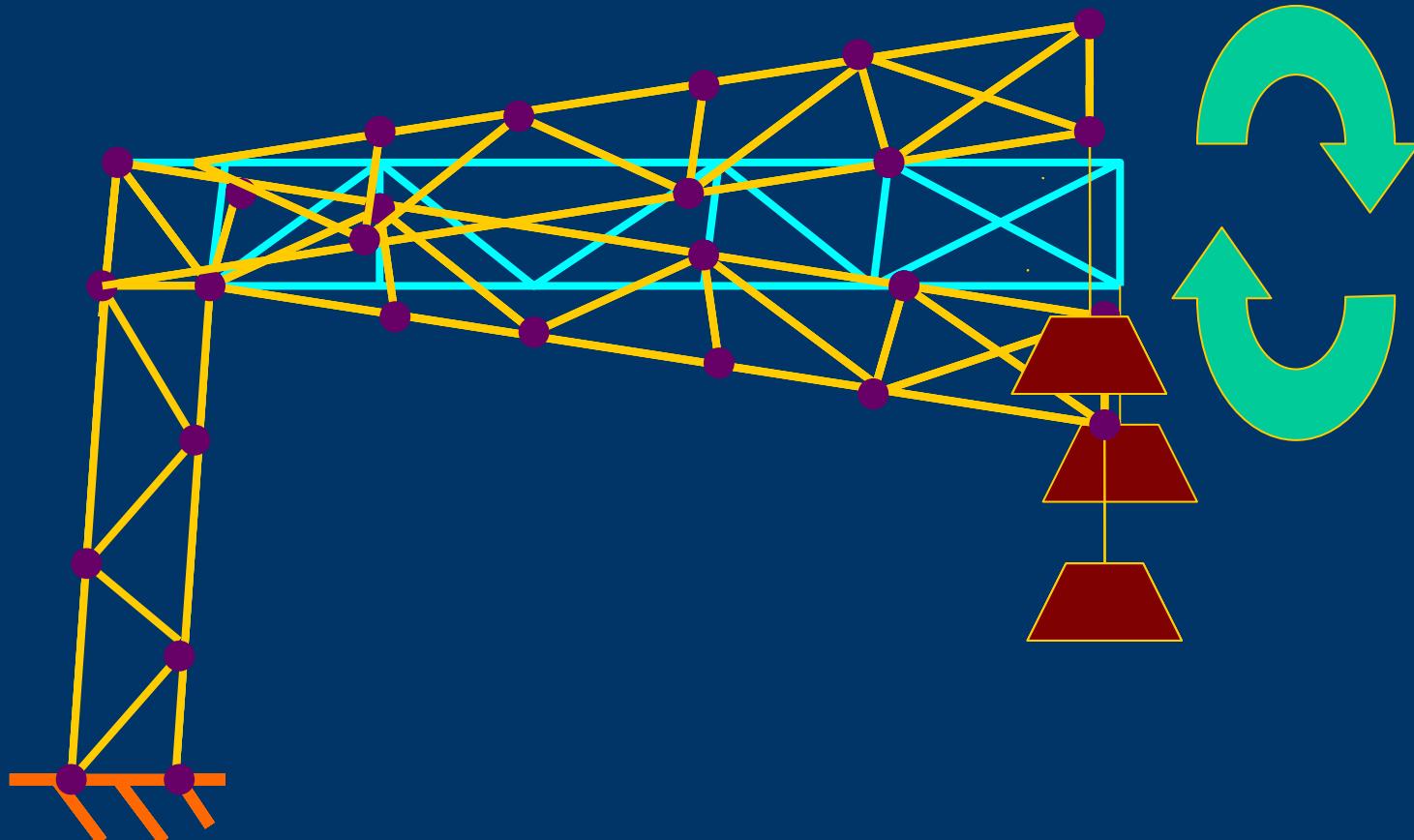


### Constructing the Model

- Cut the wire into sections.
- Model wire resistance with resistors.
- Model wire-plane capacitance with capacitors.

# Application Problems

## Oscillations in a Space Frame



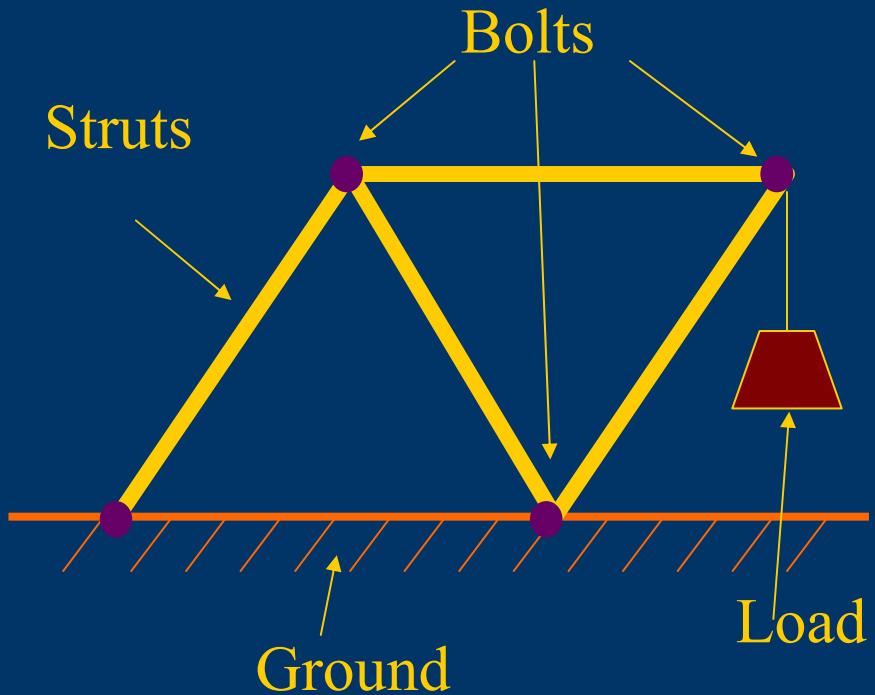
- What is the oscillation amplitude?

# Application Problems



## Oscillations in a Space Frame

### Simplified Structure

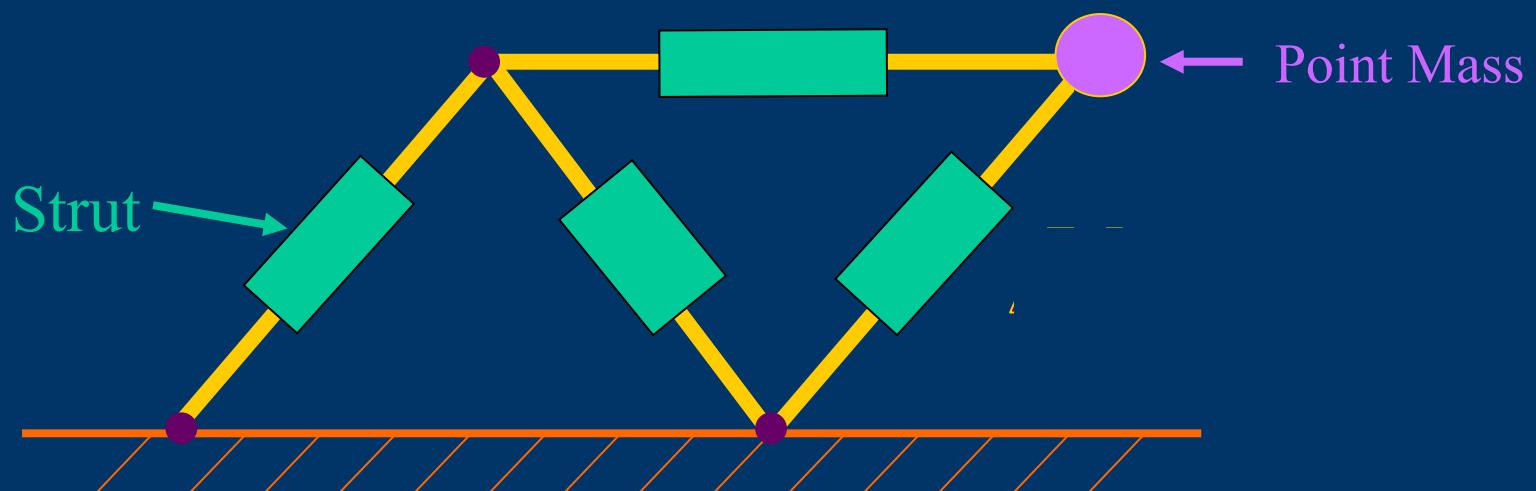


Example Simplified for Illustration

# Application Problems

## Oscillations in a Space Frame

Modeling with Struts, Joints and Point Masses

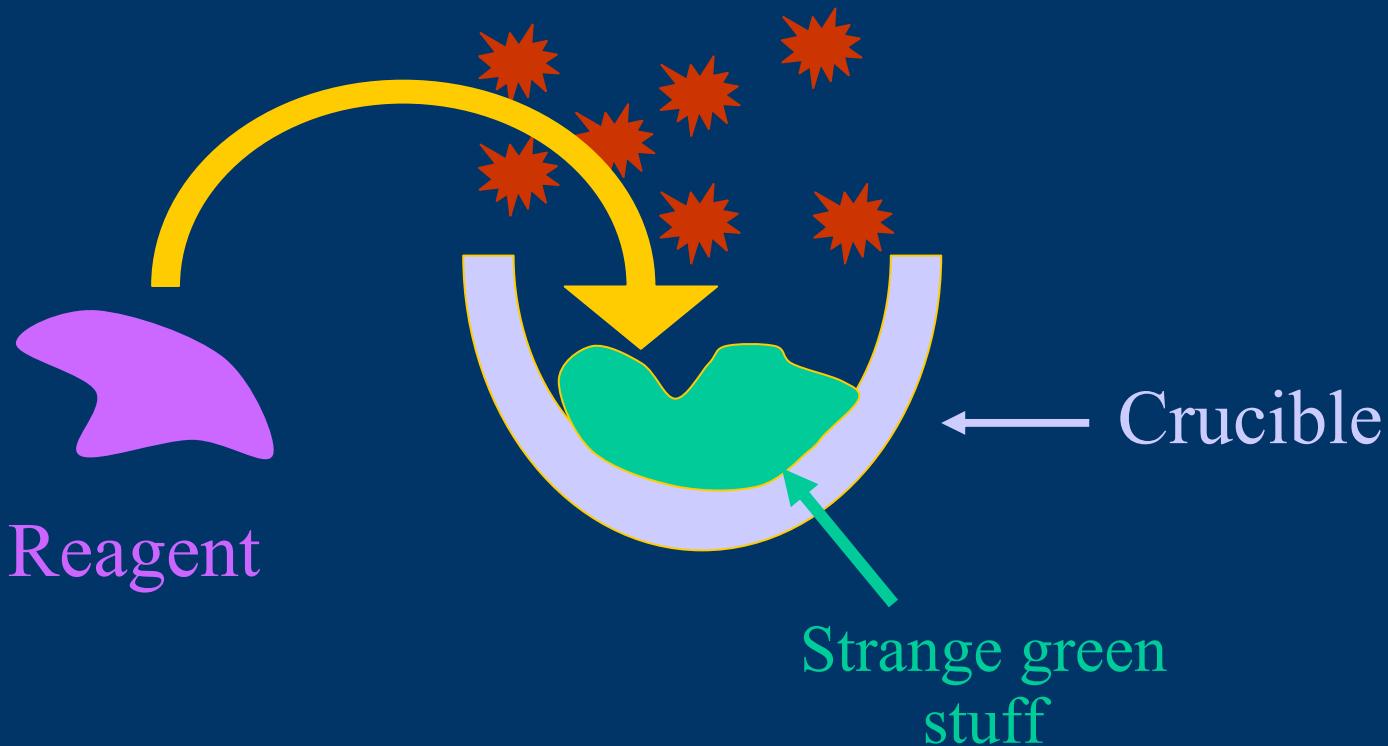


Constructing the Model

- Replace Metal Beams with Struts.
- Replace cargo with point mass.

# Application Problems

# Chemical Reaction Dynamics

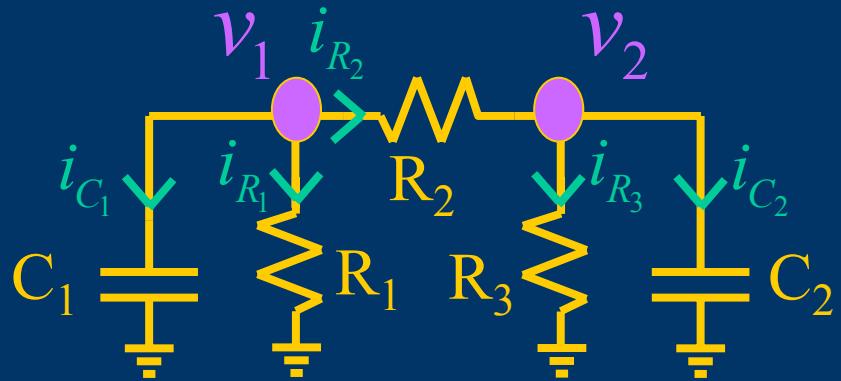


- How fast is product produced?
- Does it explode?

# Application Problems

## Signal Transmission in an Integrated Circuit

### A 2x2 Example



Constitutive Equations

$$i_c = C \frac{dv_c}{dt}$$

$$i_R = \frac{1}{R} v_R$$

Conservation Laws

$$i_{C_1} + i_{R_1} + i_{R_2} = 0$$

$$i_{C_2} + i_{R_3} - i_{R_2} = 0$$

Nodal Equations Yields 2x2 System

$$\begin{bmatrix} C_1 & 0 \\ 0 & C_2 \end{bmatrix} \begin{bmatrix} \frac{dv_1}{dt} \\ \frac{dv_2}{dt} \end{bmatrix} = - \begin{bmatrix} \frac{1}{R_1} + \frac{1}{R_2} & -\frac{1}{R_2} \\ -\frac{1}{R_2} & \frac{1}{R_3} + \frac{1}{R_2} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

# Application Problems

## Signal Transmission in an Integrated Circuit

A 2x2 Example

Let  $C_1 = C_2 = 1$ ,  $R_1 = R_3 = 10$ ,  $R_2 = 1$

$$\begin{bmatrix} C_1 & 0 \\ 0 & C_2 \end{bmatrix} \begin{bmatrix} \frac{dv_1}{dt} \\ \frac{dv_2}{dt} \end{bmatrix} = - \begin{bmatrix} \frac{1}{R_1} + \frac{1}{R_2} & -\frac{1}{R_2} \\ -\frac{1}{R_2} & \frac{1}{R_3} + \frac{1}{R_2} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix}$$

➡

$$\frac{dx}{dt} = \underbrace{\begin{bmatrix} -1.1 & 1.0 \\ 1.0 & -1.1 \end{bmatrix}}_A x$$

Eigenvalues and Eigenvectors

$$A = \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} -0.1 & 0 \\ 0 & -2.1 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 1 & 1 \end{bmatrix}^{-1}$$

eigenvectors      Eigenvalues

# An Aside on Eigenanalysis

Consider an ODE:  $\frac{dx(t)}{dt} = Ax(t), \quad x(0) = x_0$

Eigendecomposition:  $A = \underbrace{\begin{bmatrix} \vdots & \vdots & \vdots \\ E_1 & E_2 & E_n \\ \vdots & \vdots & \vdots \end{bmatrix}}_E \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} \begin{bmatrix} \vdots & \vdots & \vdots \\ E_1 & E_2 & E_n \\ \vdots & \vdots & \vdots \end{bmatrix}^{-1}$

Change of variables:  $Ey(t) = x(t) \Leftrightarrow y(t) = E^{-1}x(t)$

Substituting:  $\frac{dEy(t)}{dt} = AEy(t), \quad Ey(0) = x_0$

Multiply by  $E^{-1}$ :  $\frac{dy(t)}{dt} = E^{-1}AEy(t) = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} y(t)$

# An Aside on Eigenanalysis Continued

From last slide:  $\frac{dy(t)}{dt} = \begin{bmatrix} \lambda_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \lambda_n \end{bmatrix} y(t)$  Decoupled Equations!

Decoupling:  $\frac{dy_i(t)}{dt} = \lambda_i y_i(t) \Rightarrow y_i(t) = e^{\lambda_i t} y(0)$

Steps for solving  $\frac{dx(t)}{dt} = Ax(t), \quad x(0) = x_0$

1) Determine  $E, \lambda$

2) Compute  $y(0) = E^{-1}x_0 \begin{bmatrix} e^{\lambda_1 t} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & e^{\lambda_n t} \end{bmatrix}$

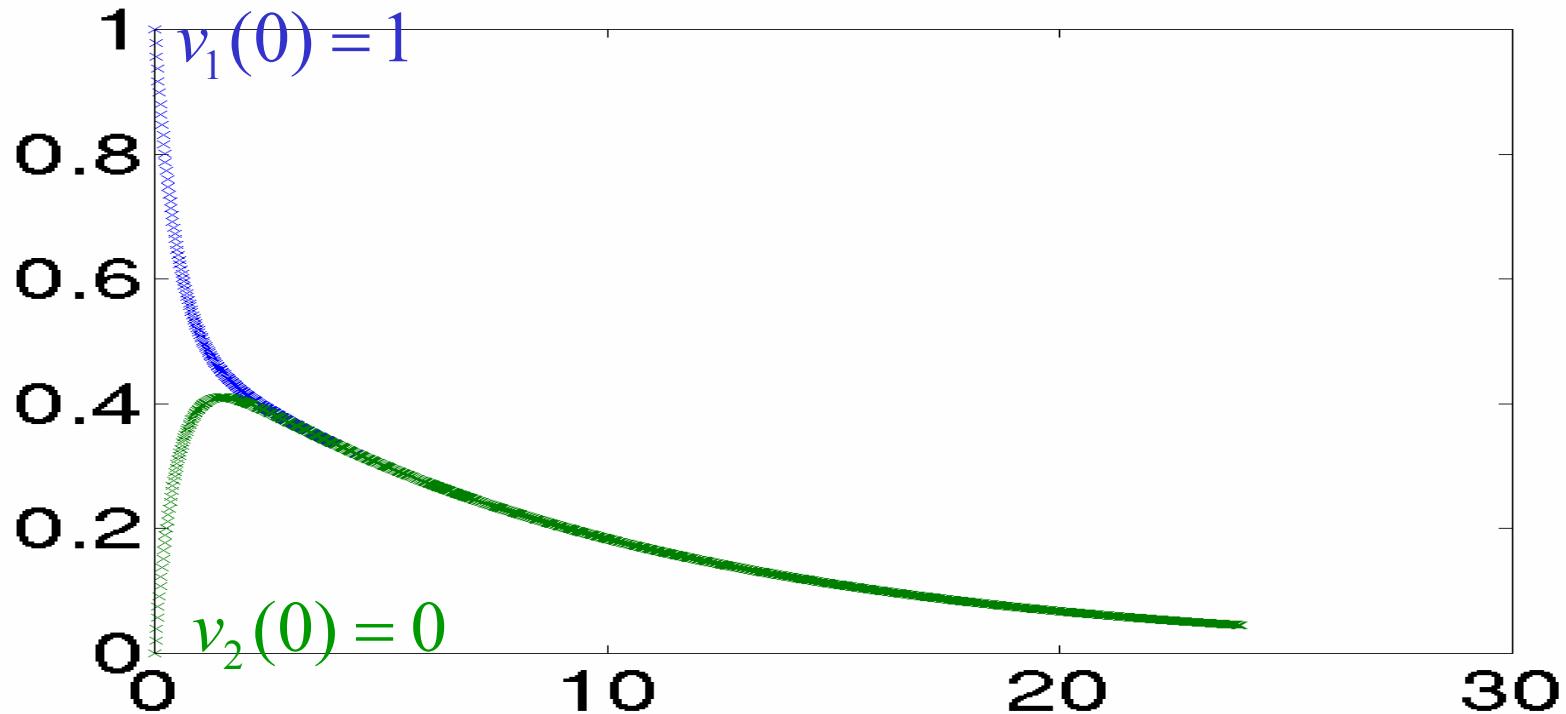
3) Compute  $y(t) = \begin{bmatrix} e^{\lambda_1 t} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & e^{\lambda_n t} \end{bmatrix} y(0)$

4)  $x(t) = Ey(t)$

## Application Problems

# Signal Transmission in an Integrated Circuit

### A 2x2 Example



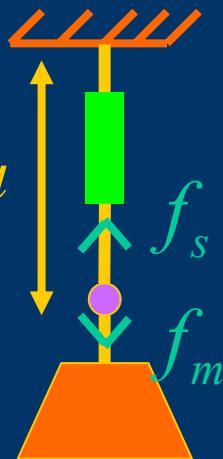
Notice two time scale behavior

- $v_1$  and  $v_2$  come together quickly (fast eigenmode).
- $v_1$  and  $v_2$  decay to zero slowly (slow eigenmode).

# Application Problems

## Struts, Joints and point mass example

### A 2x2 Example



Constitutive  
Equations

$$f_m = M \frac{d^2 u}{dt^2}$$

$$f_s = EA_c * \frac{y - y_0}{y_0} = \frac{EA_c}{y_0} u$$

Conservation  
Law

$$f_s + f_m = 0$$

Define v as velocity ( $du/dt$ ) to yield a 2x2 System

$$\begin{bmatrix} M & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{dv}{dt} \\ \frac{du}{dt} \end{bmatrix} = \begin{bmatrix} 0 & -\frac{EA_c}{y_0} \\ 1 & 0 \end{bmatrix} \begin{bmatrix} v \\ u \end{bmatrix}$$

# Application Problems

# Struts, Joints and point mass example

## A 2x2 Example

Let  $M = 1$ ,  $\frac{EA_c}{l} = 1$

$$\begin{bmatrix} M & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \frac{dv}{dt} \\ \frac{du}{dt} \end{bmatrix} = \begin{bmatrix} y_0 & 0 & 1 \\ 0 & -\frac{EA_c}{y_0} & 0 \end{bmatrix} \begin{bmatrix} v \\ u \end{bmatrix} \quad \xrightarrow{\text{Large pink arrow}} \quad \frac{dx}{dt} = \underbrace{\begin{bmatrix} 0 & -1.0 \\ 1.0 & 0 \end{bmatrix}}_A x$$

# Eigenvalues and Eigenvectors

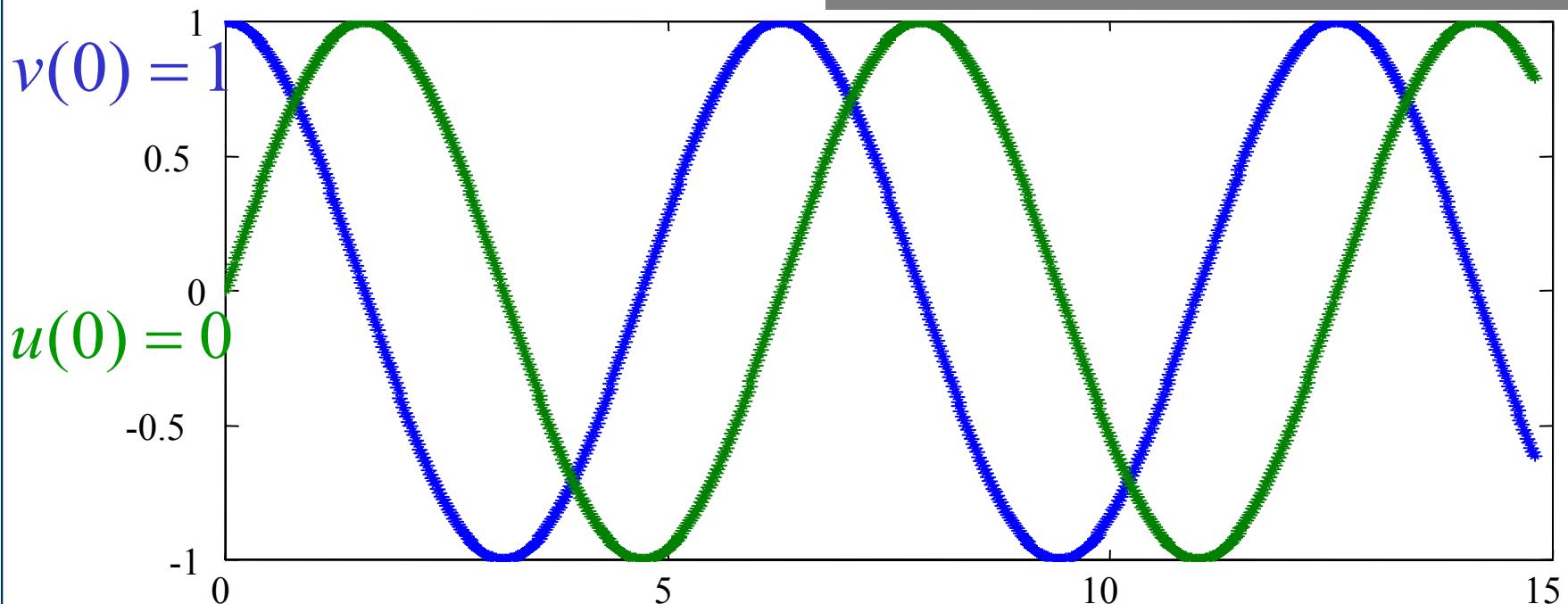
$$A = \begin{bmatrix} -1 & i \\ i & -i \end{bmatrix} \begin{bmatrix} i & 0 \\ 0 & -i \end{bmatrix} \begin{bmatrix} -1 & -1 \\ i & -i \end{bmatrix}^{-1}$$

eigenvectors                      Eigenvalues

# Application Problems

## Struts, Joints and point mass example

### A 2x2 Example



Note the system has imaginary eigenvalues

- Persistent Oscillation
- Velocity,  $v$ , peaks when displacement,  $u$ , is zero.

# Application Problems

## Chemical Reaction Example

### A 2x2 Example

Amount of reactant =  $R$ , the temperature =  $T$

$$\frac{dT}{dt} = -T + R$$

More reactant causes temperature to rise,  
higher temperatures increases heat dissipation  
causing temperature to fall

$$\frac{dR}{dt} = -R + 4T$$

Higher temperatures raises reaction rates,  
increased reactant interferes with reaction  
and slows rate.

# Application Problems

# Chemical Reaction Example

## A 2x2 Example

$$\begin{bmatrix} \frac{dT}{dt} \\ \frac{dR}{dt} \end{bmatrix} = \begin{bmatrix} -1 & 1 \\ 4 & -1 \end{bmatrix} \begin{bmatrix} T \\ R \end{bmatrix} \quad \rightarrow \quad \frac{dx}{dt} = \underbrace{\begin{bmatrix} -1 & 1 \\ 4 & -1 \end{bmatrix}}_A x$$

# Eigenvalues and Eigenvectors

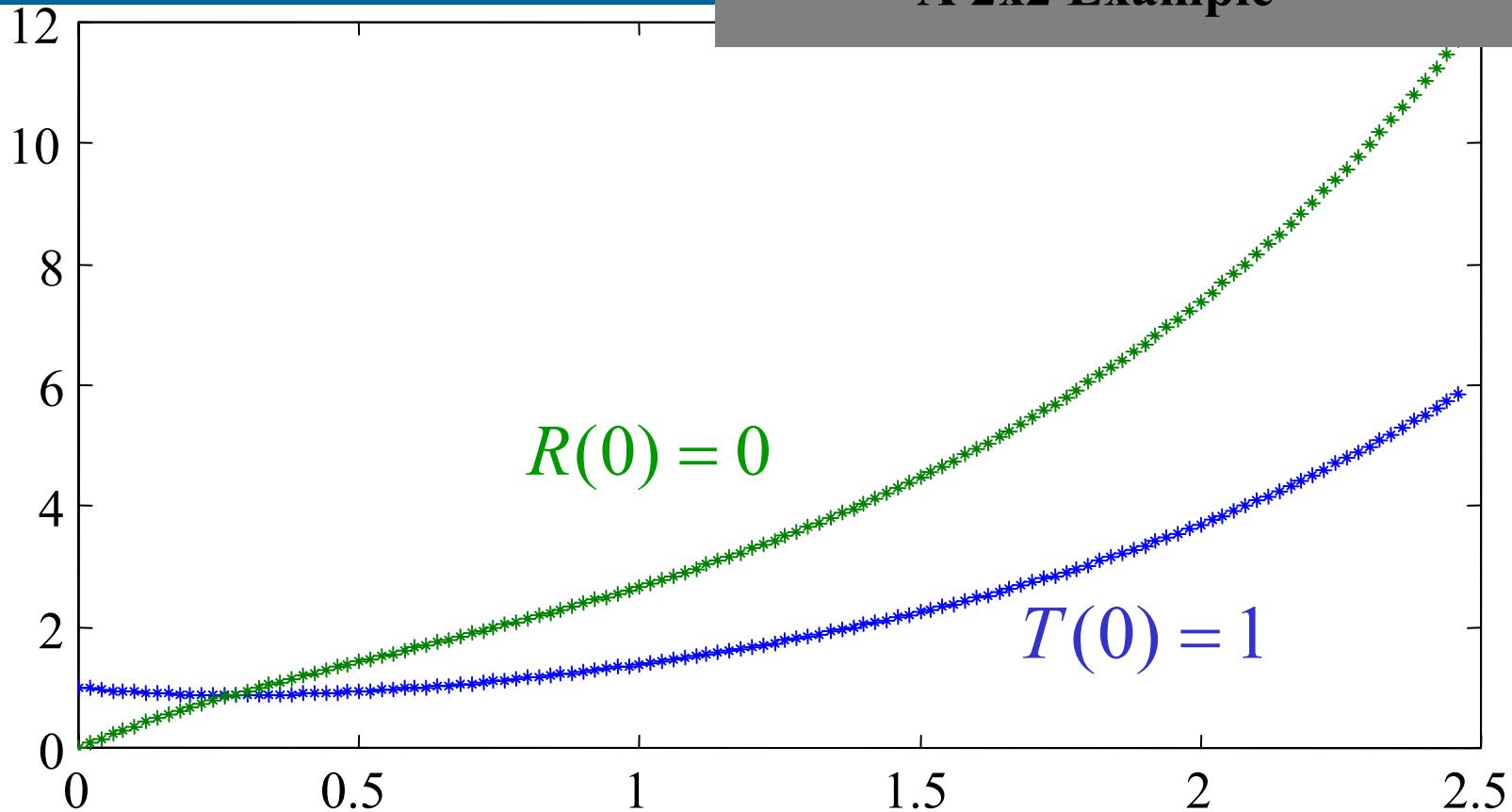
$$A = \begin{bmatrix} 1 & -1 \\ 2 & 2 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -3 \end{bmatrix} \begin{pmatrix} 1 & -1 \\ 2 & 2 \end{pmatrix}^{-1}$$

eigenvectors                      Eigenvalues

# Application Problems

## Chemical Reaction Example

A 2x2 Example

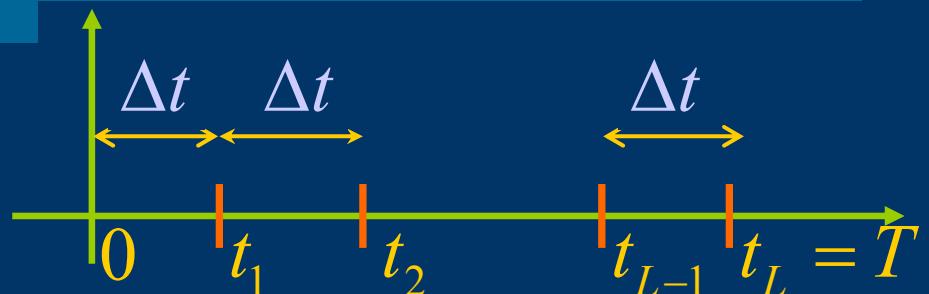


Note the system has a positive eigenvalue  
• Solutions grow exponentially with time.

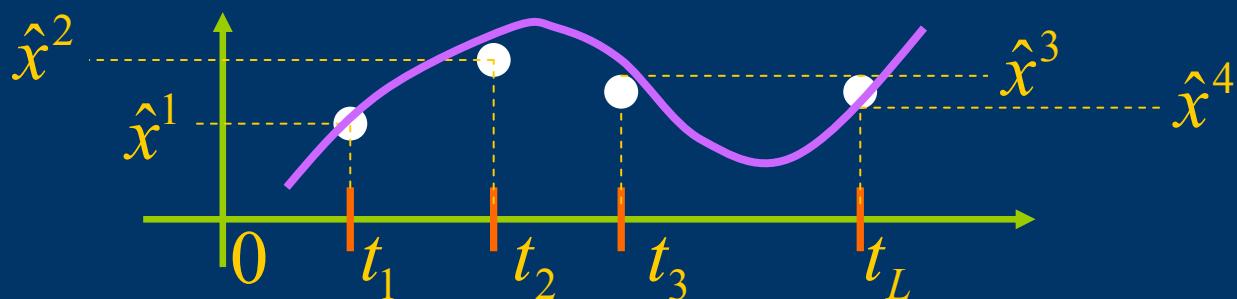
# Finite Difference Methods

## Basic Concepts

First - Discretize Time



Second - Represent  $x(t)$  using values at  $t_i$



$$\begin{array}{ll} \hat{x}^l & \simeq x(t_l) \\ \text{Approx.} & \text{Exact} \\ \text{sol'n} & \text{sol'n} \end{array}$$

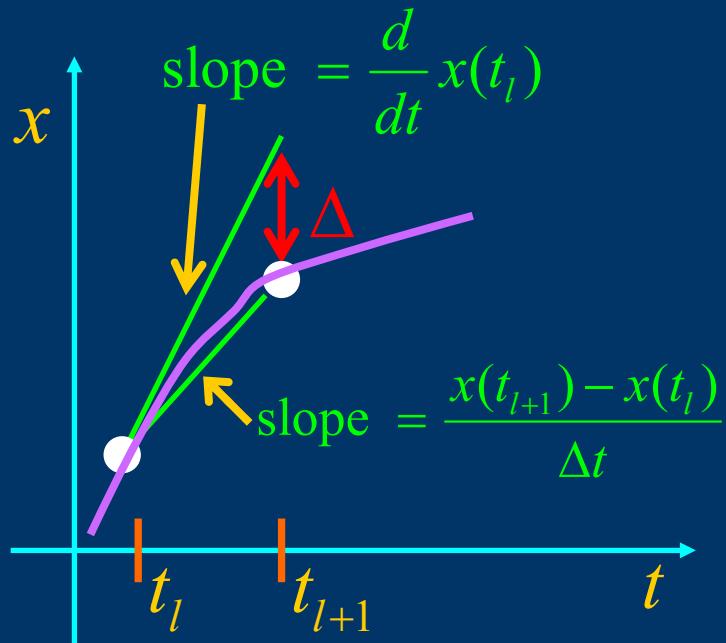
Third - Approximate  $\frac{d}{dt}x(t)$  using the discrete  $\hat{x}^l$ 's

$$\text{Example: } \frac{d}{dt}x(t_l) \simeq \frac{\hat{x}^l - \hat{x}^{l-1}}{\Delta t_l} \text{ or } \frac{\hat{x}^{l+1} - \hat{x}^l}{\Delta t_{l+1}}$$

# Finite Difference Methods

## Basic Concepts

### Forward Euler Approximation



$$\frac{dx}{dt} = A x(t_l) \cong \frac{x(t_{l+1}) - x(t_l)}{\Delta t}$$

or

$$x(t_{l+1}) \cong x(t_l) + \Delta t A x(t_l)$$

$$\Delta = x(t_{l+1}) - (x(t_l) + \Delta t A x(t_l))$$

# Finite Difference Methods

## Basic Concepts

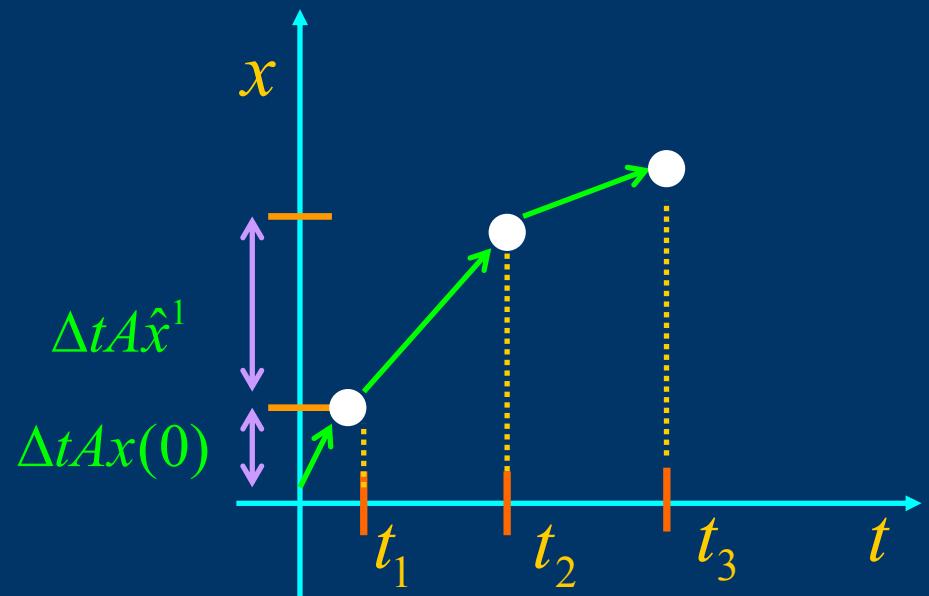
### Forward Euler Algorithm

$$x(t_1) \approx \hat{x}^1 = x(0) + \Delta t A x(0)$$

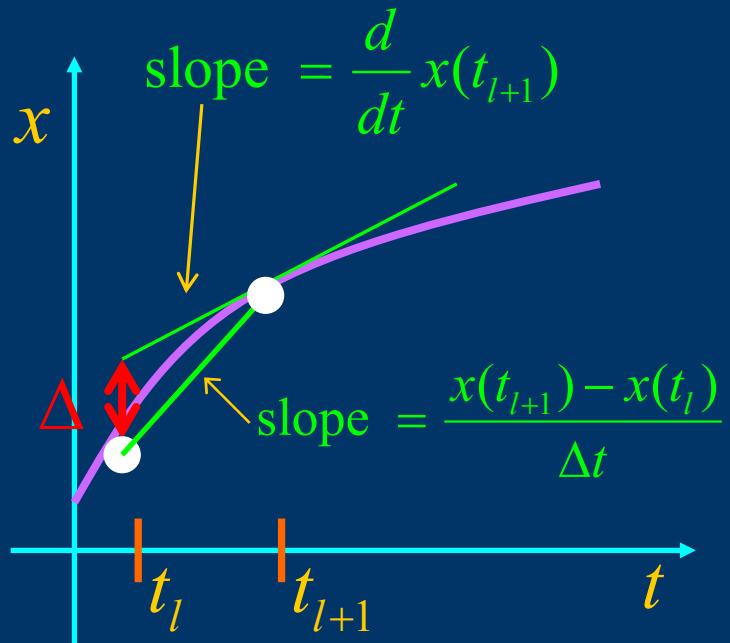
$$x(t_2) \approx \hat{x}^2 = \hat{x}^1 + \Delta t A \hat{x}^1$$

⋮

$$x(t_L) \approx \hat{x}^L = \hat{x}^{L-1} + \Delta t A \hat{x}^{L-1}$$



# Finite Difference Methods



## Basic Concepts

### Backward Euler Approximation

$$\frac{d}{dt} x(t_{l+1}) = A x(t_{l+1}) \cong \frac{x(t_{l+1}) - x(t_l)}{\Delta t}$$

or

$$x(t_{l+1}) \cong x(t_l) + \Delta t A x(t_{l+1})$$

$$\Delta = x(t_{l+1}) - (x(t_l) + \Delta t A x(t_{l+1}))$$

# Finite Difference Methods

## Basic Concepts

### Backward Euler Algorithm

Solve with Gaussian Elimination

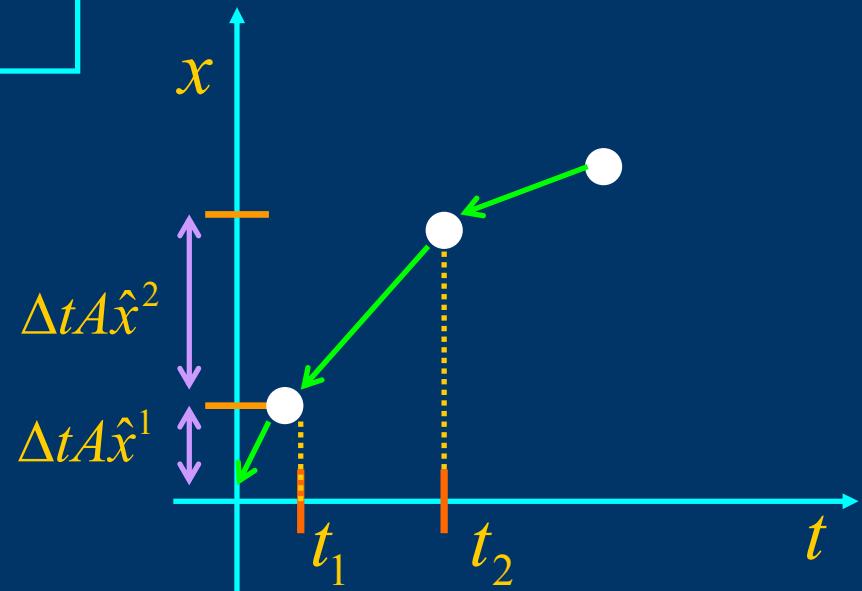
$$x(t_1) \approx \hat{x}^1 = x(0) + \Delta t A \hat{x}^1$$

$$\Rightarrow [I - \Delta t A] \hat{x}^1 = x(0)$$

$$x(t_2) \approx \hat{x}^2 = [I - \Delta t A]^{-1} \hat{x}^1$$

$\vdots$

$$x(t_L) \approx \hat{x}^L = [I - \Delta t A]^{-1} \hat{x}^{L-1}$$



# Finite Difference Methods

$$\frac{1}{2} \left( \frac{d}{dt} x(t_{l+1}) + \frac{d}{dt} x(t_l) \right)$$

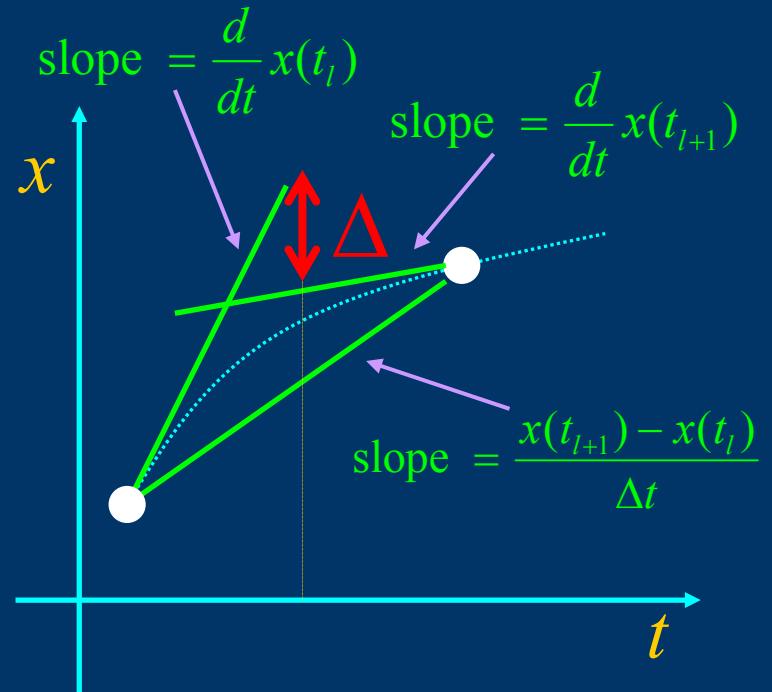
$$= \frac{1}{2} (Ax(t_{l+1}) + Ax(t_l))$$

$$\approx \frac{x(t_{l+1}) - x(t_l)}{\Delta t}$$

$$x(t_{l+1}) \approx x(t_l) + \frac{1}{2} \Delta t A(x(t_{l+1}) + x(t_l))$$

## Basic Concepts

### Trapezoidal Rule



$$\Delta = (x(t_{l+1}) - \frac{1}{2} \Delta t A x(t_l)) - (x(t_l) + \frac{1}{2} \Delta t A x(t_{l+1}))$$

# Finite Difference Methods

## Basic Concepts

### Trapezoidal Rule Algorithm

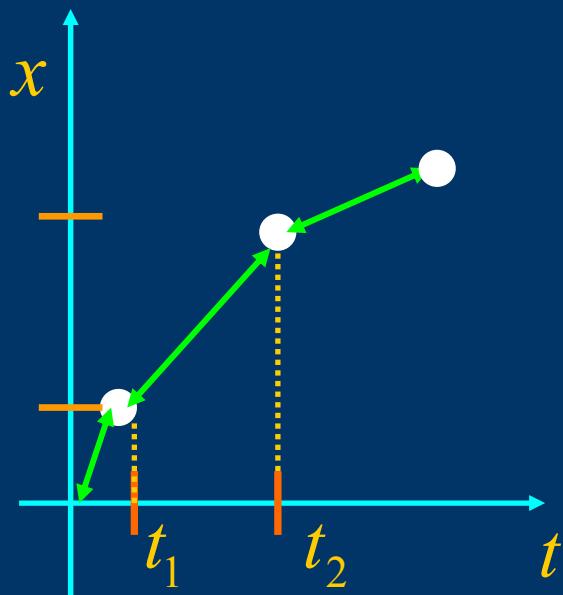
Solve with Gaussian Elimination

$$x(t_1) \approx \hat{x}^1 = x(0) + \frac{\Delta t}{2} \left( Ax(0) + A\hat{x}^1 \right)$$

$$\Rightarrow \left[ I - \frac{\Delta t}{2} A \right] \hat{x}^1 = \left[ I + \frac{\Delta t}{2} A \right] x(0)$$

$$x(t_2) \approx \hat{x}^2 = \left[ I - \frac{\Delta t}{2} A \right]^{-1} \left[ I + \frac{\Delta t}{2} A \right] \hat{x}^1$$
$$\vdots$$

$$x(t_L) \approx \hat{x}^L = \left[ I - \frac{\Delta t}{2} A \right]^{-1} \left[ I + \frac{\Delta t}{2} A \right] \hat{x}^{L-1}$$



# Finite Difference Methods

## Basic Concepts

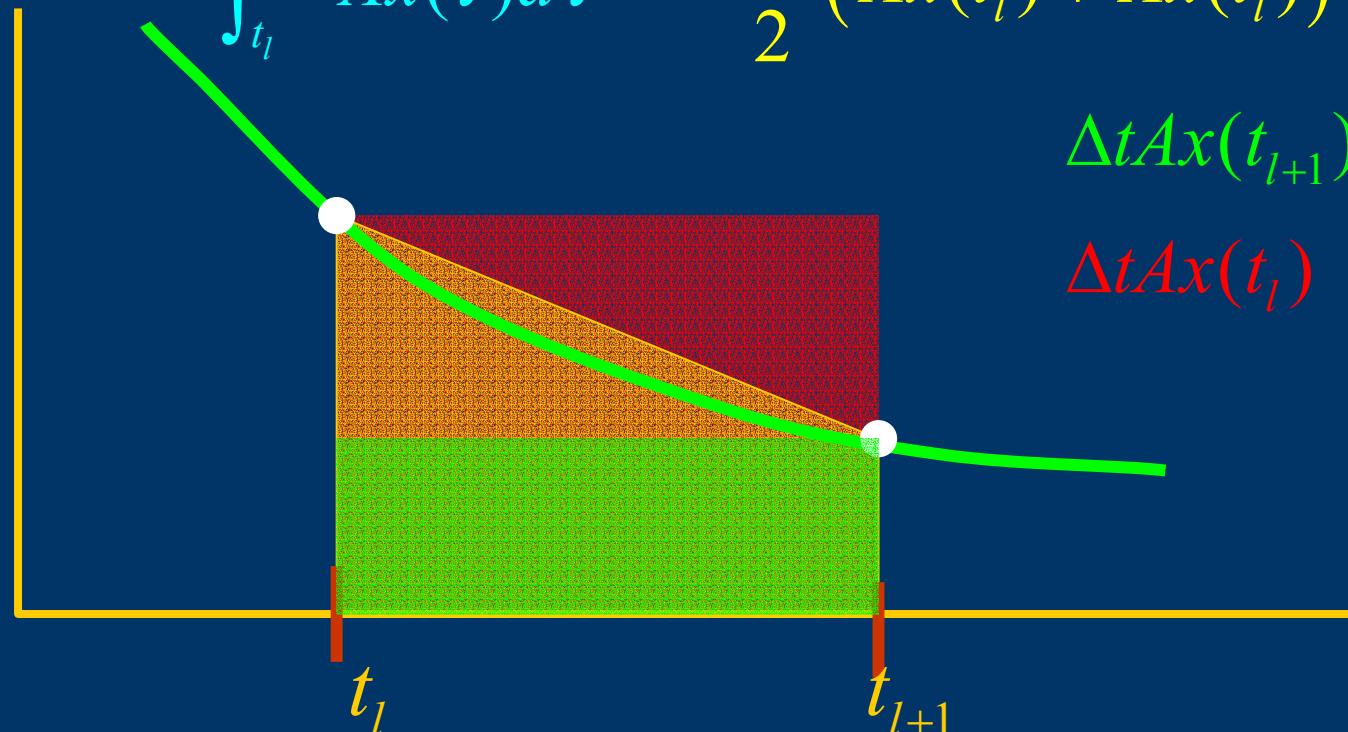
### Numerical Integration View

$$\frac{d}{dt} x(t) = Ax(t) \Rightarrow x(t_{l+1}) = x(t_l) + \int_{t_l}^{t_{l+1}} Ax(\tau)d\tau$$

$$\int_{t_l}^{t_{l+1}} Ax(\tau)d\tau \approx \frac{\Delta t}{2} (Ax(t_l) + Ax(t_{l+1})) \text{ Trap}$$

$$\Delta t Ax(t_{l+1}) \text{ BE}$$

$$\Delta t Ax(t_l) \text{ FE}$$



# Finite Difference Methods

## Basic Concepts

### Summary

Trap Rule, Forward-Euler, Backward-Euler

Are all one-step methods

$\hat{x}^l$  is computed using only  $\hat{x}^{l-1}$ , not  $\hat{x}^{l-2}, \hat{x}^{l-3}$ , etc.

Forward-Euler is simplest

No equation solution  $\rightarrow$  explicit method.

Boxcar approximation to integral

Backward-Euler is more expensive

Equation solution each step  $\rightarrow$  implicit method

Trapezoidal Rule might be more accurate

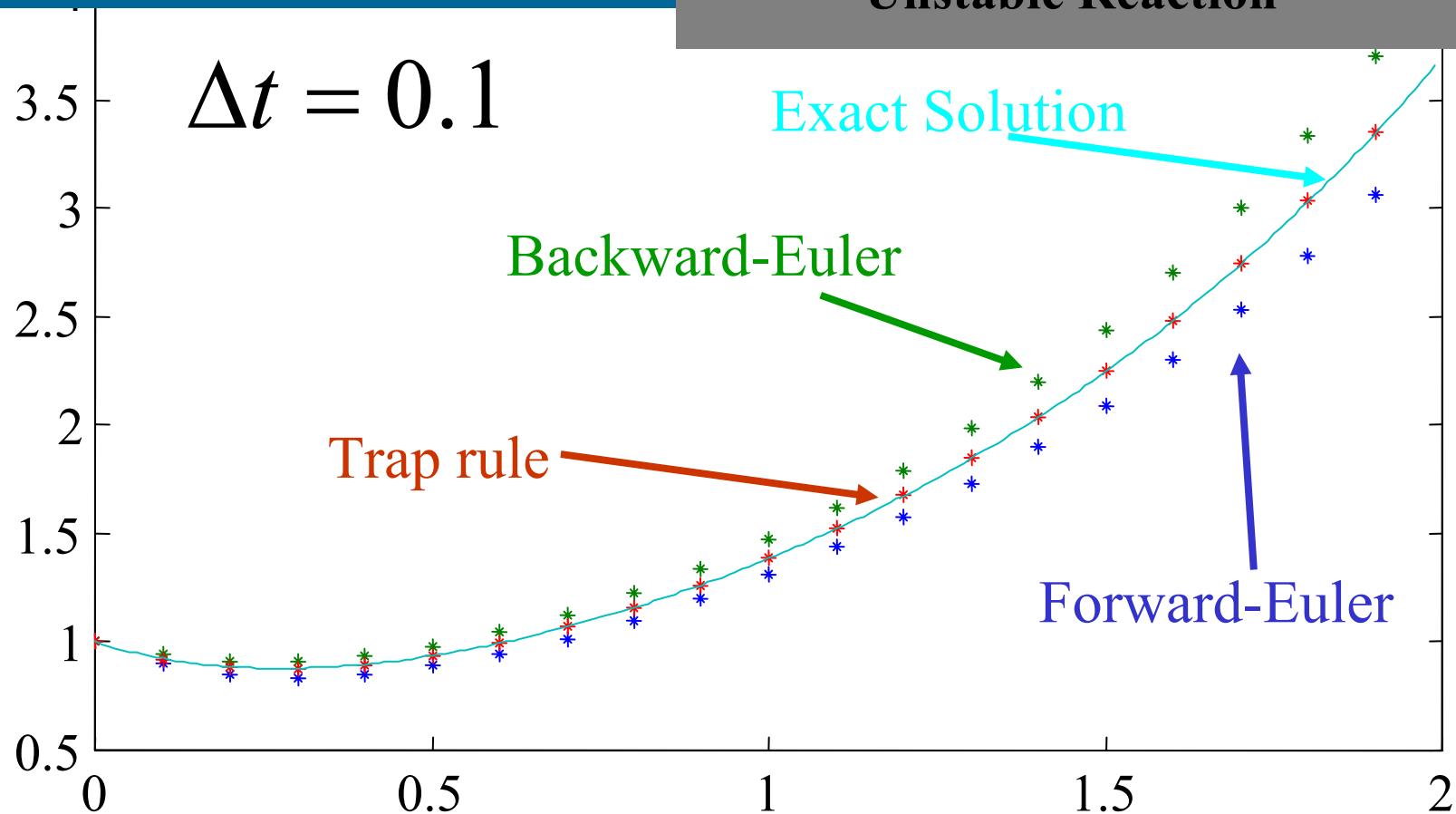
Equation solution each step  $\rightarrow$  implicit method

Trapezoidal approximation to integral

# Finite Difference Methods

## Numerical Experiments

### Unstable Reaction

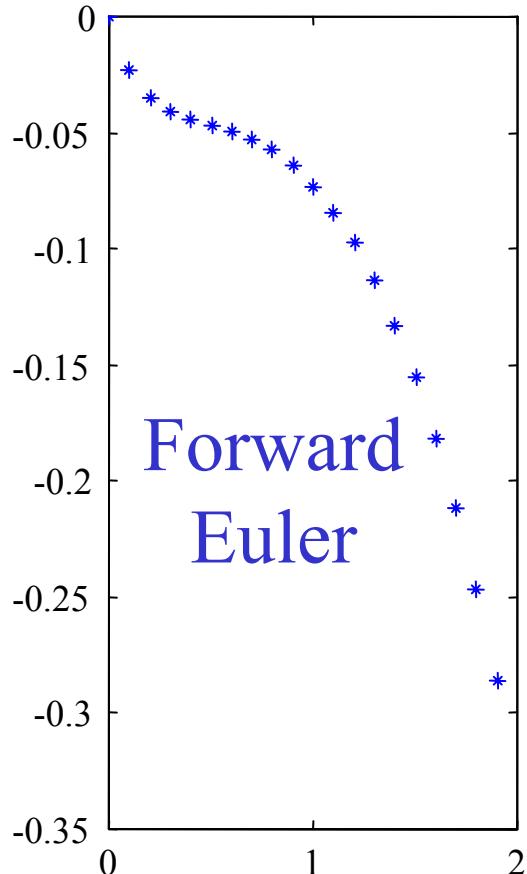


FE and BE results have larger errors than Trap Rule, and the errors grow with time.

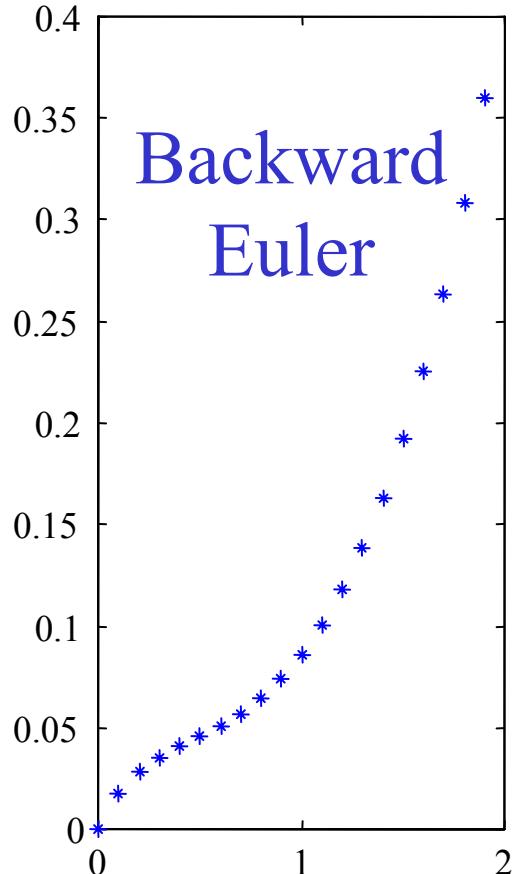
# Finite Difference Methods

## Numerical Experiments

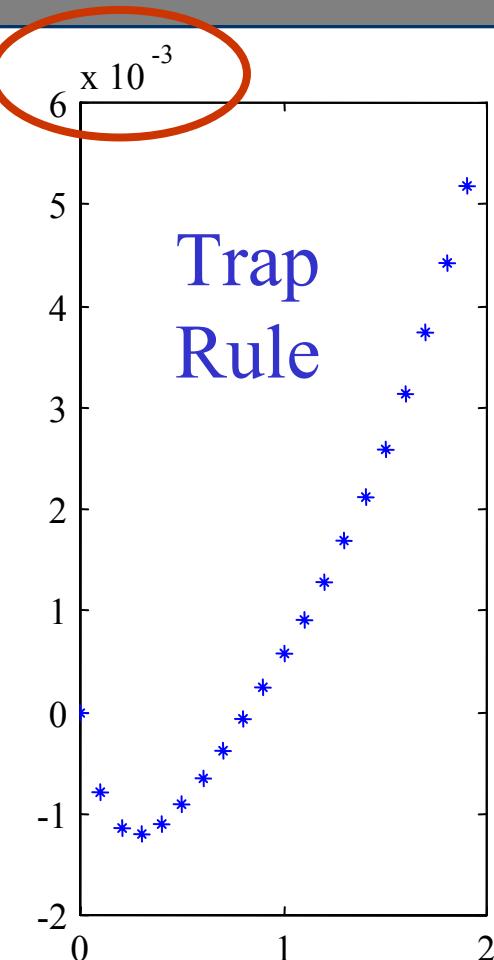
### Unstable Reaction-Error Plots



Forward Euler



Backward Euler



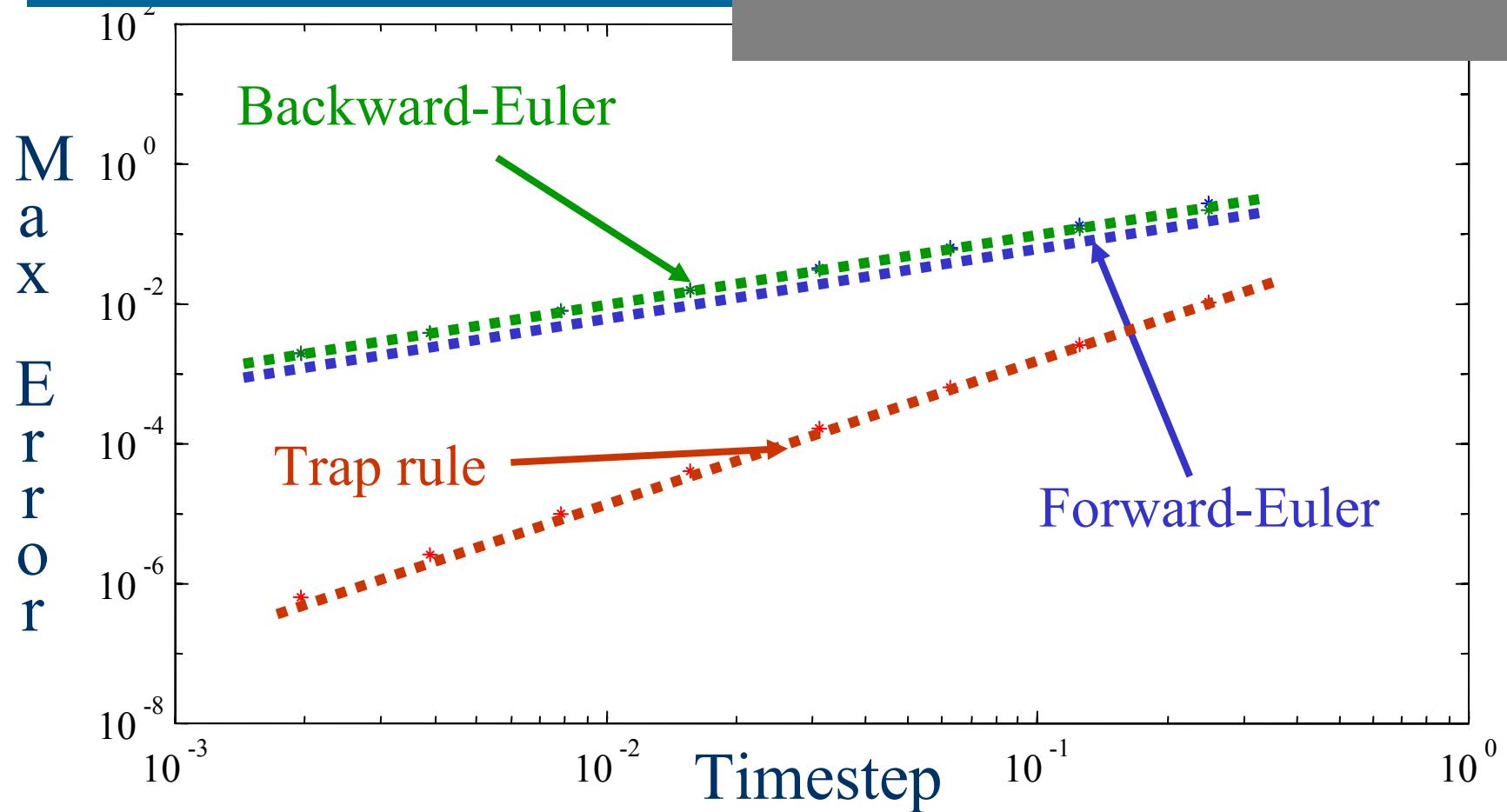
Trapezoidal Rule

All methods have errors which grow exponentially

# Finite Difference Methods

## Numerical Experiments

### Unstable Reaction-Convergence



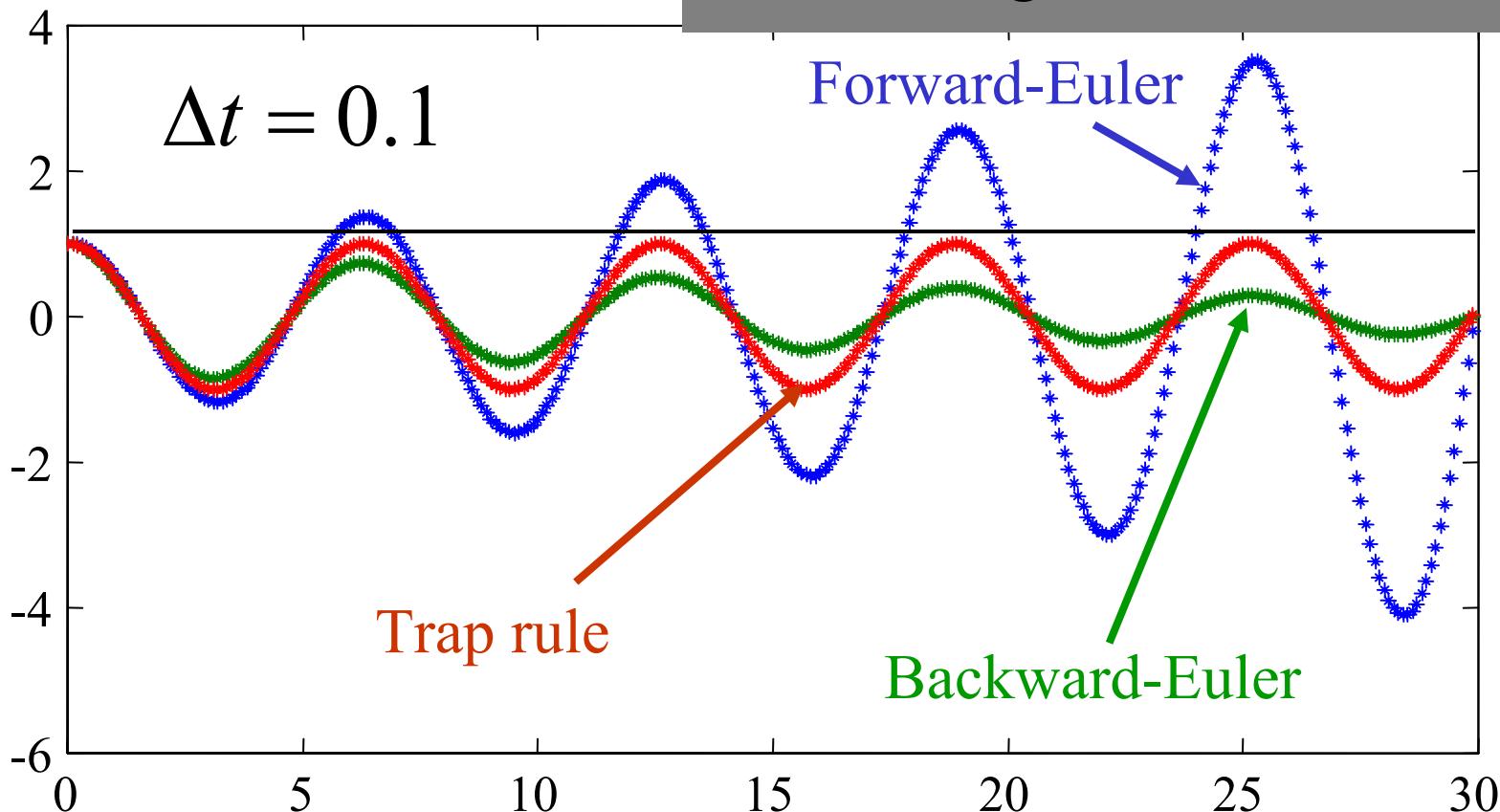
For FE and BE,  $Error \propto \Delta t$

For Trap,  $Error \propto (\Delta t)^2$

# Finite Difference Methods

## Numerical Experiments

### Oscillating Strut and Mass

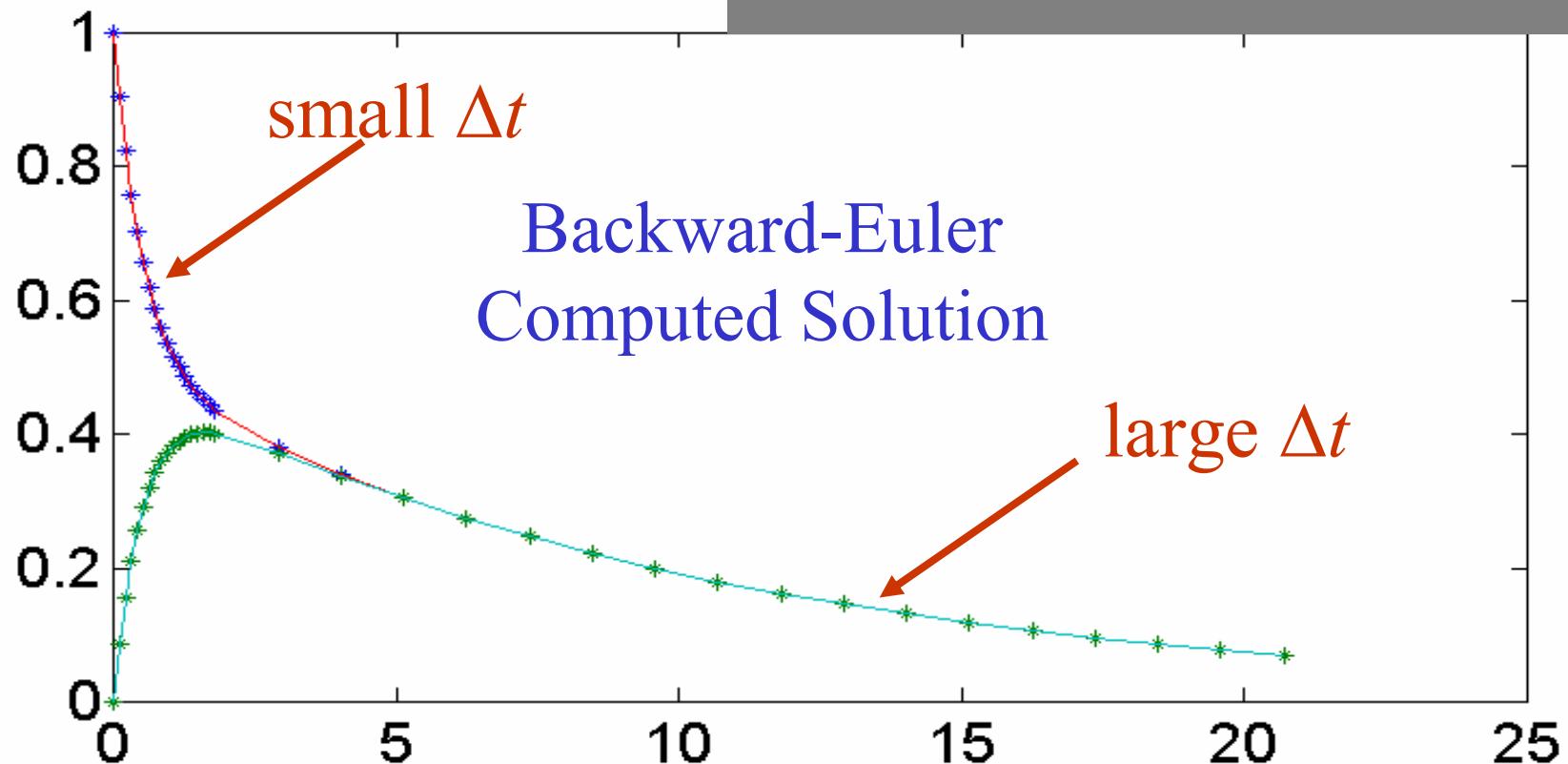


Why does FE result grow, BE result decay and the Trap rule preserve oscillations

# Finite Difference Methods

## Numerical Experiments

### Two timescale RC Circuit

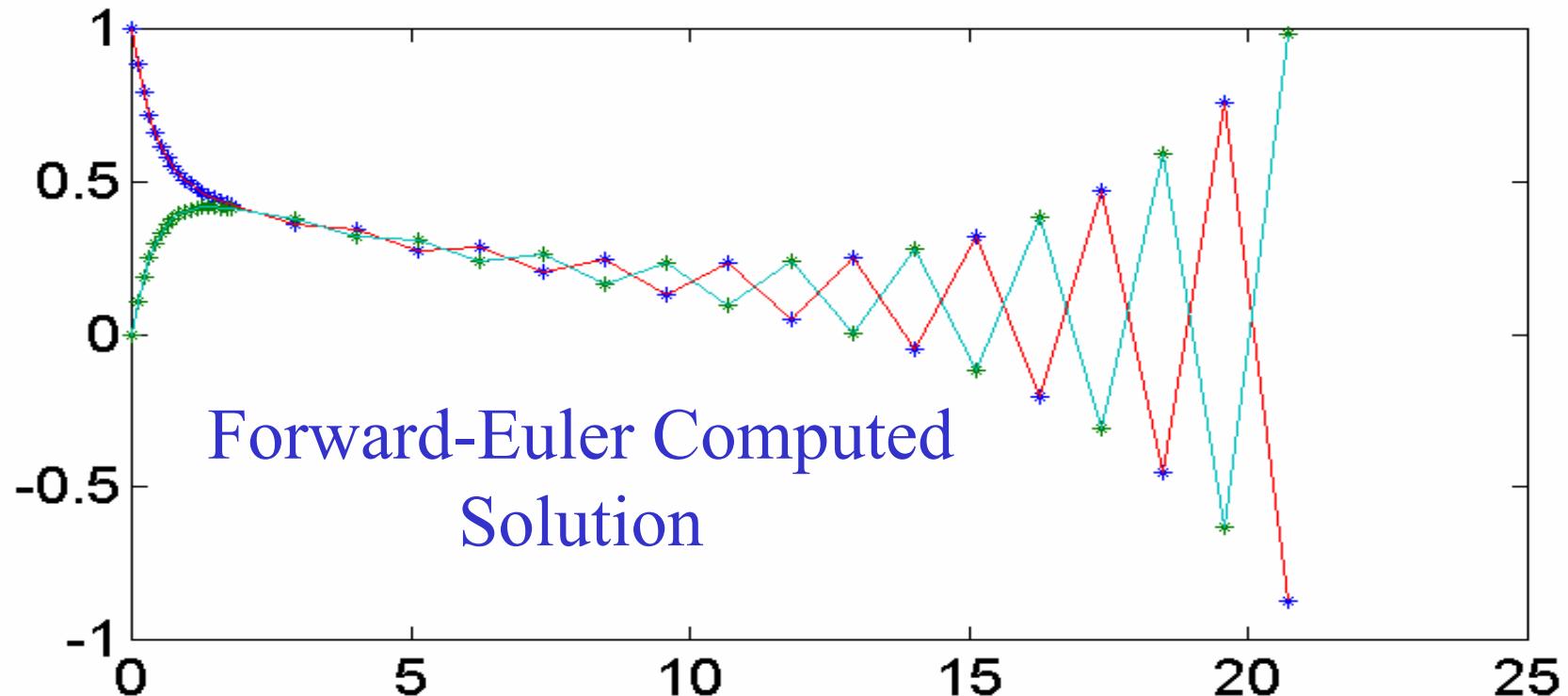


With Backward-Euler it is easy to use small timesteps for the fast dynamics and then switch to large timesteps for the slow decay

# Finite Difference Methods

## Numerical Experiments

### Two timescale RC Circuit



The Forward-Euler is accurate for small timesteps, but goes unstable when the timestep is enlarged

# Finite Difference Methods

## Numerical Experiments

### Summary

- Convergence
  - Did the computed solution approach the exact solution?
  - Why did the trap rule approach faster than BE or FE?
- Energy Preservation
  - Why did BE produce a decaying oscillation?
  - Why did FE produce a growing oscillation?
  - Why did trap rule maintain oscillation amplitude?
- Two timeconstant (stiff) problems
  - Why did FE go unstable when the timestep increased?

We will focus on convergence today

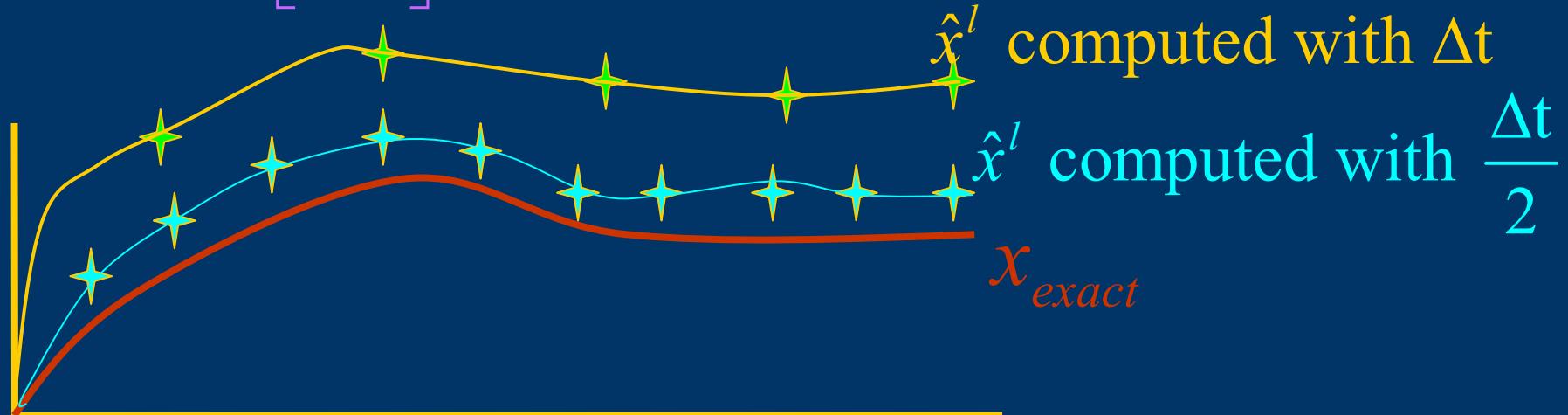
# Finite Difference Methods

## Convergence Analysis

### Convergence Definition

Definition: A finite-difference method for solving initial value problems on  $[0, T]$  is said to be convergent if given any  $A$  and any initial condition

$$\max_{l \in \left[0, \frac{T}{\Delta t}\right]} \|\hat{x}^l - x(l\Delta t)\| \rightarrow 0 \text{ as } \Delta t \rightarrow 0$$



### Order-p convergence

Definition: A finite-difference method for solving initial value problems on  $[0, T]$  is said to be order  $p$  convergent if given any  $A$  and any initial condition

$$\max_{l \in \left[0, \frac{T}{\Delta t}\right]} \|\hat{x}^l - x(l\Delta t)\| \leq C(\Delta t)^p$$

for all  $\Delta t$  less than a given  $\Delta t_0$

Forward- and Backward-Euler are order 1 convergent  
Trapezoidal Rule is order 2 convergent

# Finite Difference Methods

## Convergence Analysis

### Two Conditions for Convergence

- 1) Local Condition: One step errors are small (consistency)

Typically verified using Taylor Series

- 2) Global Condition: The single step errors do not grow too quickly (stability)

All one-step methods are stable in this sense.

# Finite Difference Methods

## Convergence Analysis

### Consistency Definition

Definition: A one-step method for solving initial value problems on an interval  $[0, T]$  is said to be consistent if for any  $A$  and any initial condition

$$\frac{\|\hat{x}^1 - x(\Delta t)\|}{\Delta t} \rightarrow 0 \text{ as } \Delta t \rightarrow 0$$

Forward-Euler definition

$$\hat{x}^1 = x(0) + \Delta t A x(0) \quad \tau \in [0, \Delta t]$$

Expanding in  $t$  about zero yields

$$x(\Delta t) = x(0) + \Delta t \frac{dx(0)}{dt} + \frac{(\Delta t)^2}{2} \frac{d^2 x(\tau)}{dt^2}$$

Noting that  $\frac{d}{dt} x(0) = Ax(0)$  and subtracting

$$\|\hat{x}^1 - x(\Delta t)\| \leq \frac{(\Delta t)^2}{2} \left\| \frac{d^2 x(\tau)}{dt^2} \right\|$$

Proves the theorem if derivatives of  $x$  are bounded

Forward-Euler definition

$$\hat{x}^{l+1} = \hat{x}^l + \Delta t A \hat{x}^l$$

Expanding in  $t$  about  $l\Delta t$  yields

$$x((l+1)\Delta t) = x(l\Delta t) + \Delta t A x(l\Delta t) + e^l$$

where  $e^l$  is the "one-step" error bounded by

$$e^l \leq C(\Delta t)^2, \text{ where } C = 0.5 \max_{\tau \in [0, T]} \left\| \frac{d^2 x(\tau)}{dt^2} \right\|$$

# Finite Difference Methods

## Convergence Analysis

### Convergence Analysis for Forward Euler Continued

Subtracting the previous slide equations

$$\hat{x}^{l+1} - x((l+1)\Delta t) = (I + \Delta t A)(\hat{x}^l - x(l\Delta t)) + e^l$$

Define the "Global" error  $E^l \equiv x^l - \hat{x}(l\Delta t)$

$$E^{l+1} = (I + \Delta t A)E^l + e^l$$

Taking norms and using the bound on  $e^l$

$$\begin{aligned}\|E^{l+1}\| &\leq \|(I + \Delta t A)\| \|E^l\| + C(\Delta t)^2 \\ &\leq (1 + \Delta t \|A\|) \|E^l\| + C(\Delta t)^2\end{aligned}$$

# Finite Difference Methods

## Convergence Analysis

A helpful bound on difference equations

A lemma bounding difference equation solutions

If  $|u^{l+1}| \leq (1 + \varepsilon) |u^l| + b, \quad u^0 = 0, \quad \varepsilon > 0$

Then  $|u^l| \leq \frac{e^{\varepsilon l}}{\varepsilon} |b|$

To prove, first write  $u^l$  as a power series and sum

$$|u^l| \leq \sum_{j=0}^{l-1} (1 + \varepsilon)^j |b| = \frac{1 - (1 + \varepsilon)^l}{1 - (1 + \varepsilon)} |b|$$

# Finite Difference Methods

## Convergence Analysis

A helpful bound on difference  
equations cont.

To finish, note  $(1 + \varepsilon) \leq e^\varepsilon \Rightarrow (1 + \varepsilon)^l \leq e^{\varepsilon l}$

$$|u^l| \leq \frac{1 - (1 + \varepsilon)^j}{1 - (1 + \varepsilon)} |b| = \frac{(1 + \varepsilon)^j - 1}{\varepsilon} |b| \leq \frac{e^{\varepsilon l}}{\varepsilon} |b|$$

Mapping the global error equation to the lemma

$$\|E^{l+1}\| \leq \left( 1 + \underbrace{\Delta t \|A\|}_{\varepsilon} \right) \|E^l\| + \underbrace{C(\Delta t)^2}_b$$

# Finite Difference Methods

## Convergence Analysis

Back to Forward Euler  
Convergence analysis.

Applying the lemma and cancelling terms

$$\|E^l\| \leq \left( 1 + \underbrace{\Delta t \|A\|}_{\varepsilon} \right) \|E^{l-1}\| + \underbrace{C(\Delta t)^2}_{b} \leq \frac{e^{l\Delta t \|A\|}}{\cancel{\Delta t \|A\|}} C(\Delta t)^2$$

Finally noting that  $l\Delta t \leq T$ ,

$$\max_{l \in [0, L]} \|E^l\| \leq e^{\|A\|T} \frac{C}{\|A\|} \Delta t$$

# Finite Difference Methods

## Convergence Analysis

Observations about the forward-Euler analysis.

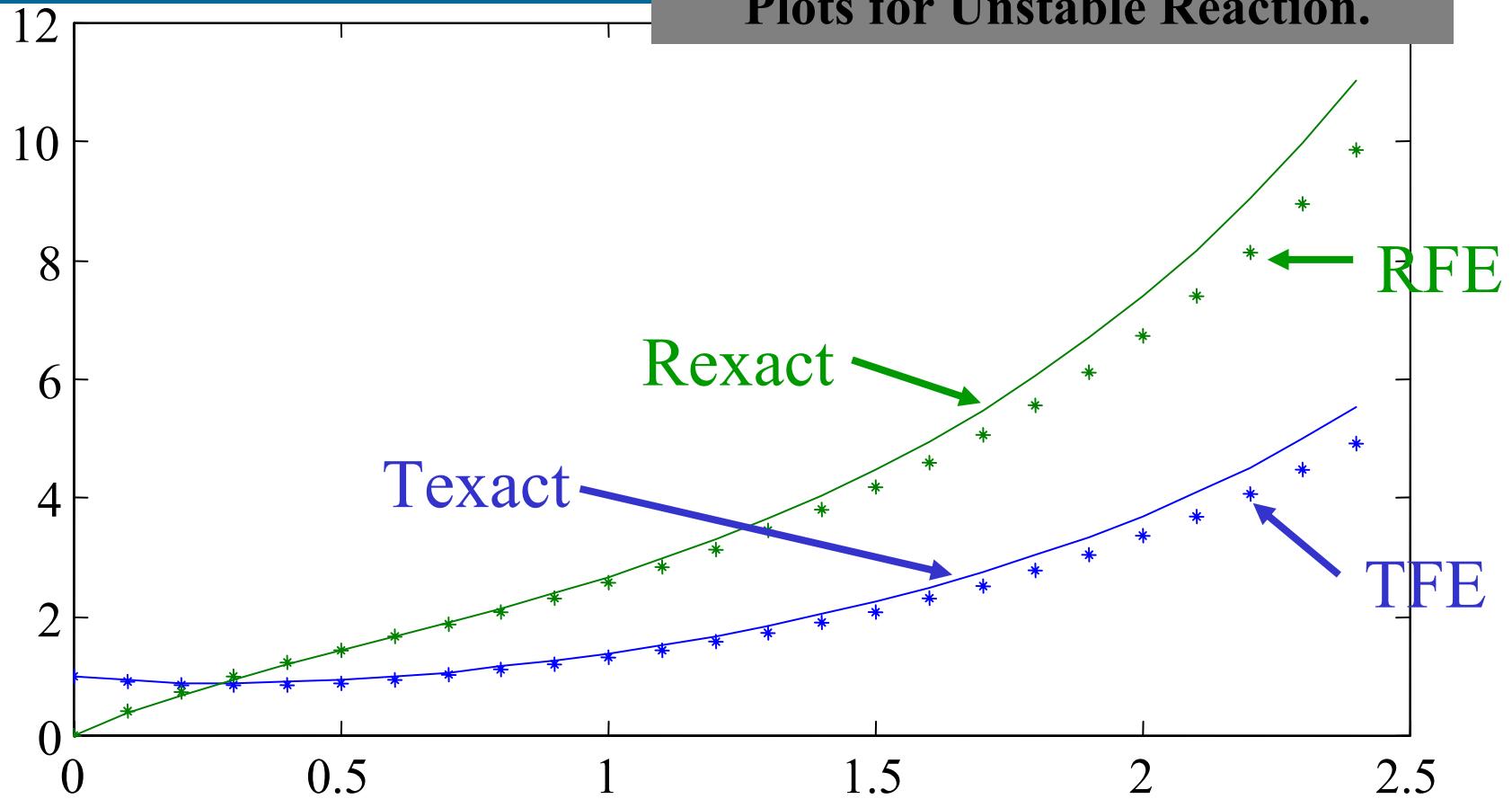
$$\max_{l \in [0, L]} \|E^l\| \leq e^{\|A\|T} \frac{C}{\|A\|} \Delta t$$

- forward-Euler is order 1 convergent
- The bound grows exponentially with time interval
- C is related to the solution second derivative
- The bound grows exponentially fast with  $\text{norm}(A)$ .

# Finite Difference Methods

## Convergence Analysis

Exact and forward-Euler(FE)  
Plots for Unstable Reaction.

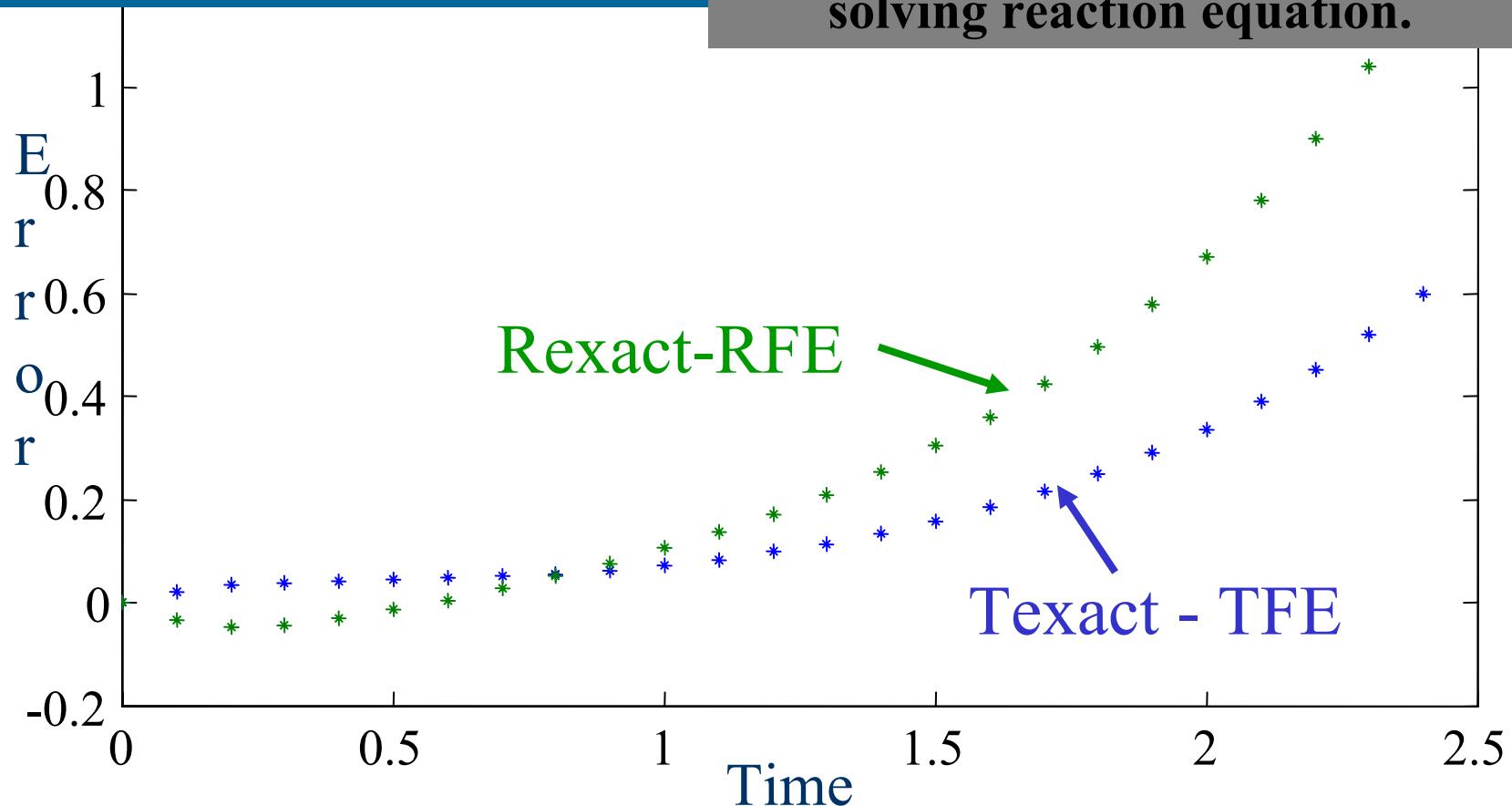


Forward-Euler Errors appear to grow with time

# Finite Difference Methods

## Convergence Analysis

forward-Euler errors for solving reaction equation.

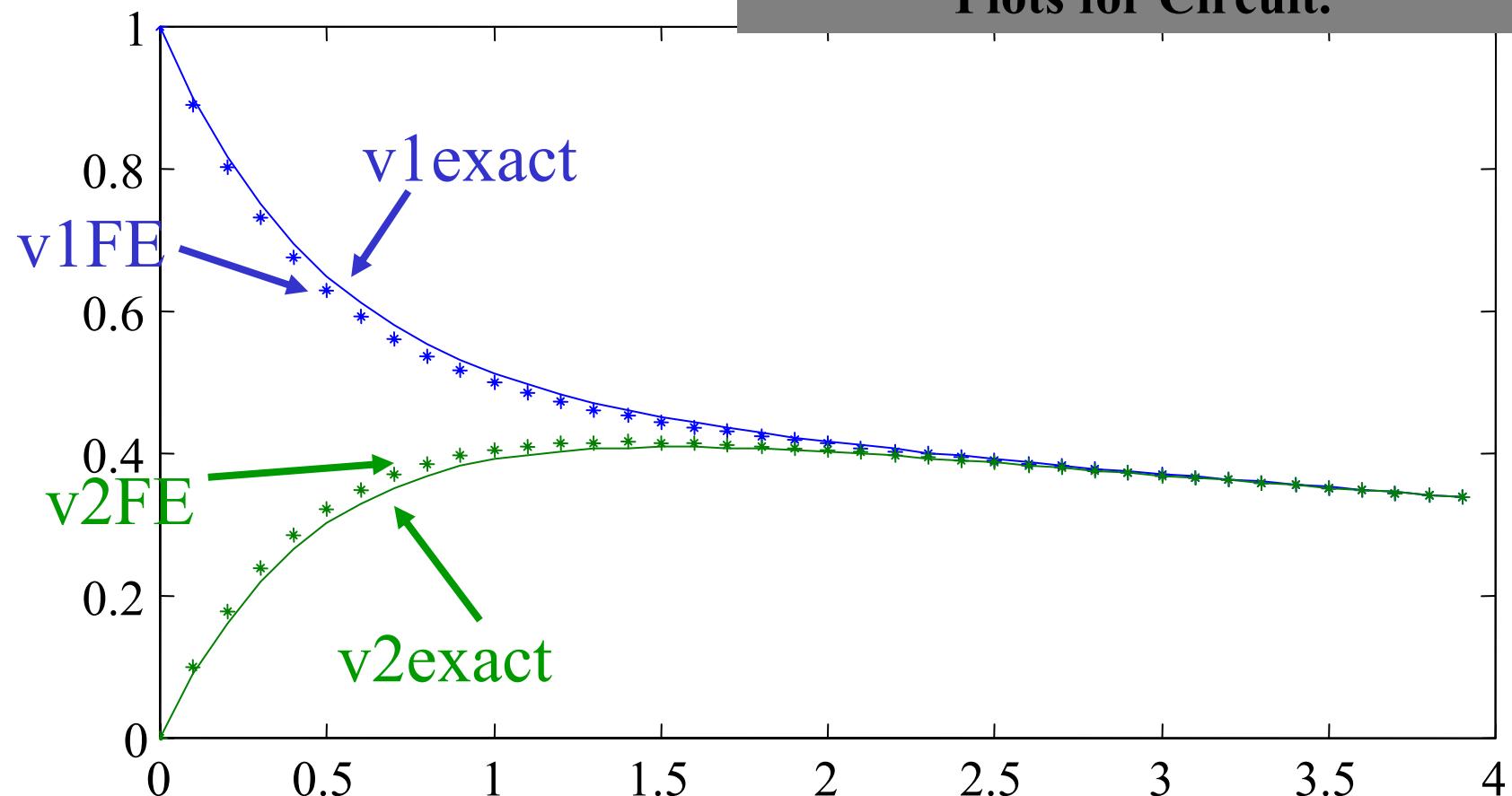


Note error grows exponentially with time, as bound predicts

# Finite Difference Methods

## Convergence Analysis

Exact and forward-Euler(FE)  
Plots for Circuit.

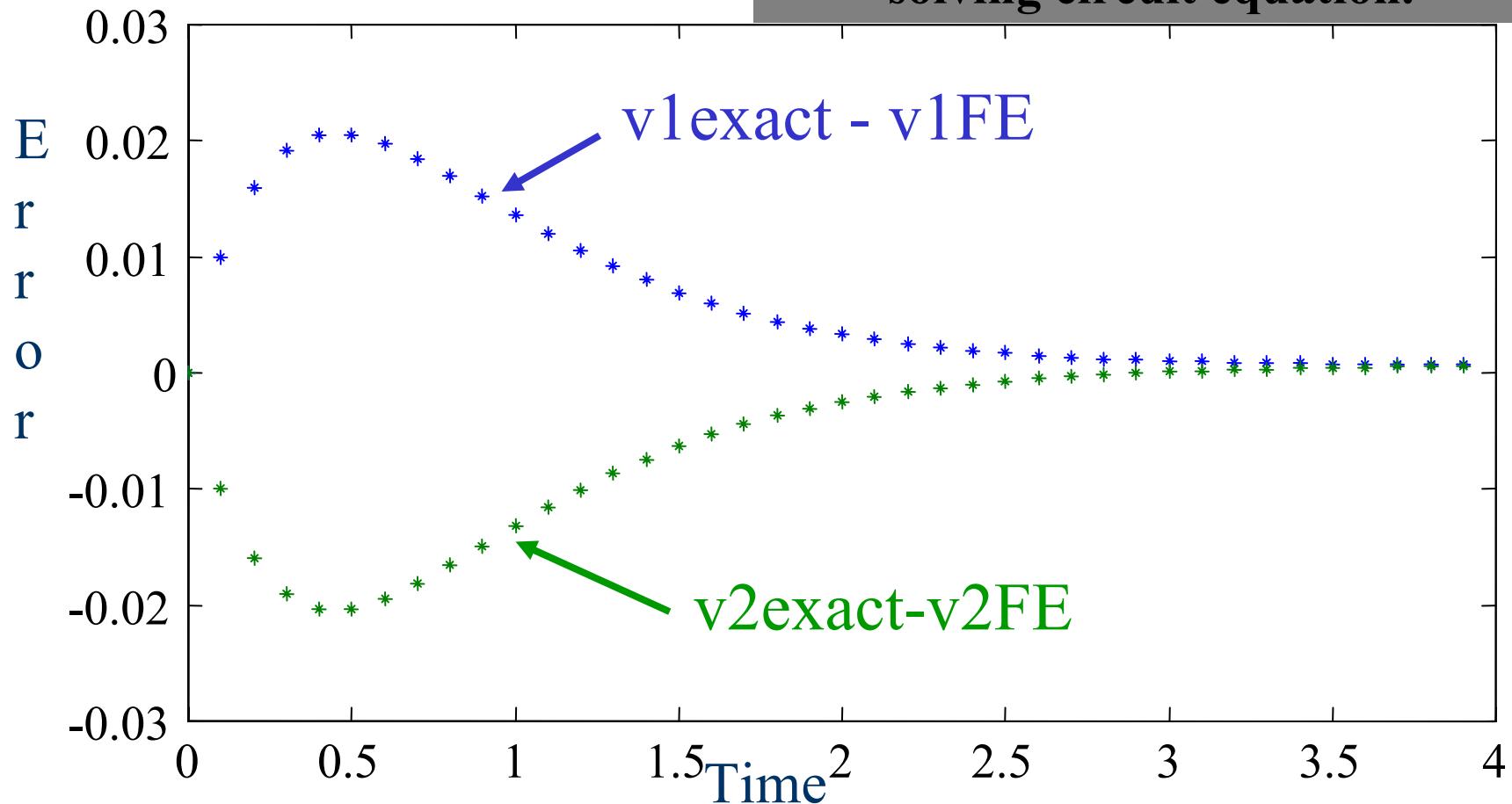


Forward-Euler Errors don't always grow with time

# Finite Difference Methods

## Convergence Analysis

forward-Euler errors for solving circuit equation.



Error does not always grow exponentially with time!  
Bound is conservative

# Summary

Initial Value problem examples

Signal propagation (two time scales).

Space frame dynamics (oscillator).

Chemical reaction dynamics (unstable system).

Looked at the simple finite-difference methods

Forward-Euler, Backward-Euler, Trap Rule.

Look at the approximations and algorithms

Experiments generated many questions

Analyzed Convergence for Forward-Euler

**Many more questions to answer, some next time**