

# NUMERICAL SOLUTION OF ODE IVPs

## Overview

1. Quick review of direction fields.
2. A reminder about \_\_\_\_\_(1) and \_\_\_\_\_(2).
3. Important test: Is the ODE initial value problem \_\_\_\_\_(3)?
4. Fundamental concepts: Euler's Method.
5. Fundamental concepts: Truncation error.
6. Fundamental concepts: \_\_\_\_\_(4) of a method.
7. Fundamental concepts: \_\_\_\_\_(5) of a method.
8. Stiff ODEs.
9. Other methods overview.
10. Systems and higher order IVPs.
11. Solving IVPs with packaged software.

# Numerical Solution of ODE IVPs

## Direction Field Review

General First Order Ordinary Differential Equation:

$$y' = f(t, y)$$

- $y'$  is shorthand for \_\_\_\_\_(1).
- $f(t, y)$  is a function of the \_\_\_\_\_(2) variable  $t$  and the \_\_\_\_\_(3) variable  $y$ .

Assumptions:

- $f(t, y)$  is defined and single valued in some rectangular region  $R$  in the  $t - y$  plane.
- If  $y = y(t)$  is a solution, then it is differentiable at all points in  $R$ . This allows us to plot a smooth curve.

# Numerical Solution of ODE IVPs

## Direction Field Review – Demo

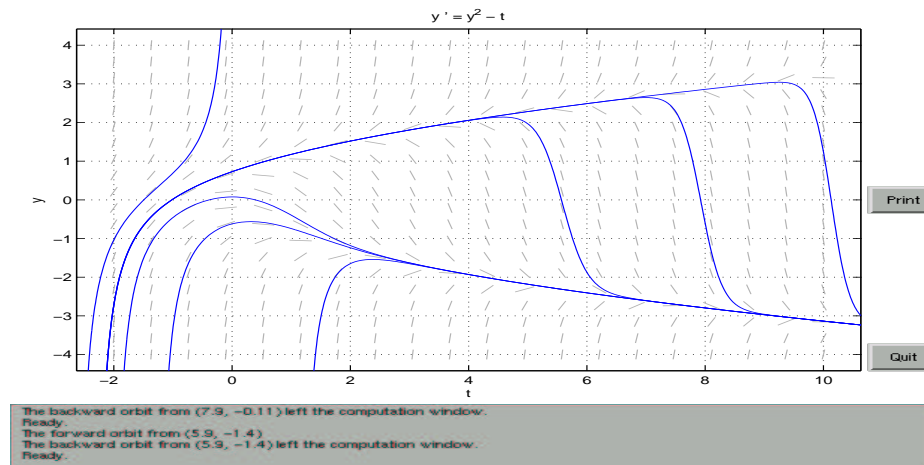
A direction field should be plotted \_\_\_\_\_(1).

**Step 1:** Draw a region in the  $t - y$  plane.

**Step 2:** Choose a point  $(a, b)$  in the region.

**Step 3:** Plot a short line starting at  $(a, b)$  with slope  $f(a, b)$ .

**Step 4:** Repeat steps (2) and (3) for many different points  $(a, b)$ .



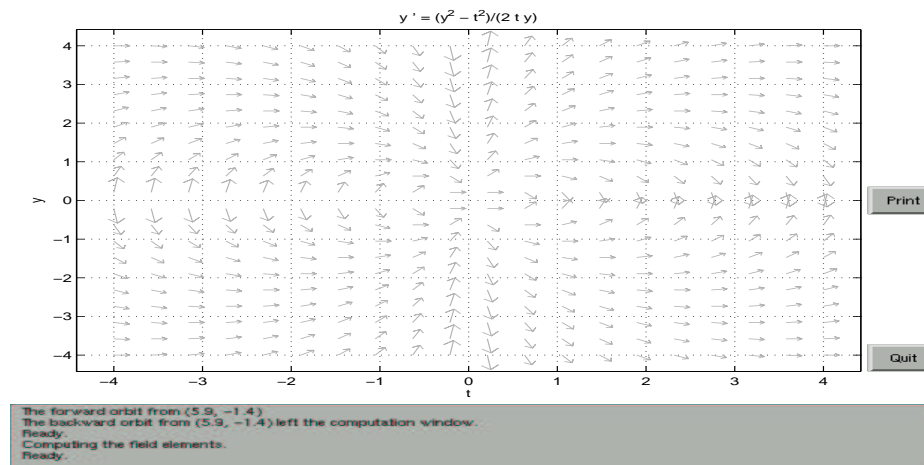
# Numerical Solution of ODE IVPs

## Direction Field Review – Notes

- A direction field gives a sense of the \_\_\_\_\_(1) of the solutions.
- Warning: be careful plotting a line that has a \_\_\_\_\_(2) (dividing by zero).

Example:

$$y' = (y^2 - t^2)/(2ty)$$



# Numerical Solution of ODE IVPs

## Existence and Uniqueness: A Reminder

Questions we must ask:

- Is there a solution?
- Is there only one solution?

Why are these questions important?

1. If there is \_\_\_\_\_(1), your computer program may \_\_\_\_\_(2).
2. If there is \_\_\_\_\_(3), the program will \_\_\_\_\_(4). It may not be the one you want.

From your ODEs course, you learned theorems that gave checklists ensuring the existence and uniqueness of ODE IVP solutions.

**A TIP:** Use these theorems.

## Numerical Solution of ODE IVPs

### Well-Posed Problems: A Must for Computation

Before you begin computing a solution to an IVP, you must ensure that

- the problem is \_\_\_\_\_(1)

This automatically ensures that the solution to the problem

- exists
- is unique

**Basic meaning:** The solution of a *well-posed problem* is not only unique, but also is

\_\_\_\_\_ (2) to

\_\_\_\_\_ (3)

in the data (which almost always will occur on a computer).

# Numerical Solution of ODE IVPs

## Well-Posed Problems: Formal Definition

**Definition:** The IVP

$$y' = f(t, y), \quad a \leq t \leq b, \quad y(a) = y_0$$

is a *well-posed* problem if

(1) A \_\_\_\_\_(1) solution  $y(t)$  to the problem \_\_\_\_\_(2).

(2) A number  $\epsilon > 0$  exists such that a \_\_\_\_\_(3) solution  $z(t)$  to the  
\_\_\_\_\_ (4)

$$z' = f(t, z) + \delta(t), \quad a \leq t \leq b, \quad z(a) = y_0 + \epsilon_0$$

exists whenever \_\_\_\_\_(5) and \_\_\_\_\_(6)

(3) A constant  $\kappa > 0$  exists with the property that

$$|z(t) - y(t)| < \kappa\epsilon \quad \text{for all } t \in [a, b]$$

## Numerical Solution of ODE IVPs

### Well-Posed Problems: An Easy Test?

**Question:** How can we tell whether criteria (1) – (3) for well-posedness are satisfied for a particular problem?

**Answer:** Good news! There is an *easy test* (i.e., a theorem) that tells us immediately whether an IVP is well-posed.



## Numerical Solution of ODE IVPs

### Well-Posed Problems: An Easy Test!

**Theorem:** Suppose  $D = \{(t, y) | t \in [a, b] \text{ and } y \in [c, d]\}$ . The IVP

$$\frac{dy}{dt} = f(t, y), \quad t \in [a, b], \quad y(a) = y_0$$

is **well-posed** provided

1.  $f$  is \_\_\_\_\_(1) on  $D$
2.  $f$  satisfies a \_\_\_\_\_(2) in the variable  
\_\_\_\_\_ (3) on the set \_\_\_\_\_(4)

## Numerical Solution of ODE IVPs

### Lipschitz Continuity: Definition

**Definition:** A function  $f(t, y)$  is said to be *Lipschitz continuous* or to satisfy a *Lipschitz condition* in the variable  $y$  on a set  $D \subset \mathbb{R}^2$  provided a constant  $L > 0$  exists with the property that

$$|f(t, y_1) - f(t, y_2)| \leq L|y_1 - y_2|$$

whenever  $(t, y_1), (t, y_2) \in D$ . The constant  $L$  is called the *Lipschitz constant*.

**Note:** If  $f(t, y)$  is differentiable, then the Lipschitz condition guarantees that

\_\_\_\_\_ (1). Conversely, if  $f$  is differentiable with respect to  $y$  and

\_\_\_\_\_ (2), then  $f$  satisfies the Lipschitz condition. This property can

be used as a \_\_\_\_\_ (3) of whether the \_\_\_\_\_ (4) is satisfied.

## Numerical Solution of ODE IVPs

### Lipschitz Condition Test: Example

**Example:** Determine whether the IVP  $y' = f(t, y)$ ,  $y(t_0) = y_0$  on  $D$  is well posed, given

$$f(t, y) = ty$$

and

$$D = \{(t, y) | t \in [1, 2], y \in [-3, 4]\}$$

If so, find the Lipschitz constant.

**Answer:** For each  $(t, y_1), (t, y_2)$  in  $D$ , we have

$$|f(t, y_1) - f(t, y_2)| = |ty_1 - ty_2| = t|y_1 - y_2| \leq 2|y_1 - y_2|$$

So,  $f(t, y)$  is \_\_\_\_\_(1) in \_\_\_\_\_(2) on  
\_\_\_\_\_(3), with Lipschitz constant  $L =$  \_\_\_\_\_(4).

## Numerical Solution of ODE IVPs

### Well-Posed versus Stable Solutions (1)

**Question:** How much can a perturbed solution  $\hat{\vec{y}}(t)$  with perturbed  $\hat{\vec{f}}(t, \hat{\vec{y}})$  and perturbed initial conditions  $\hat{\vec{y}}(t_0) = \hat{\vec{y}}_0$  \_\_\_\_\_(1) from the original solution  $\vec{y}(t)$  with original  $\vec{f}(t, \vec{y})$  and original initial conditions  $\vec{y}(t_0) = \vec{y}_0$  when \_\_\_\_\_(2) continuity with constant  $L$  on a bounded domain  $D$  is assumed?

**Answer:**

$$\|\hat{\vec{y}}(t) - \vec{y}(t)\| \leq e^{L(t-t_0)} \|\hat{\vec{y}}_0 - \vec{y}_0\| + \frac{e^{L(t-t_0)} - 1}{L} \|\hat{\vec{f}} - \vec{f}\|$$

where  $\|\hat{\vec{f}} - \vec{f}\| = \max_{(t, \vec{y}) \in D} \|\hat{\vec{f}}(t, \hat{\vec{y}}) - \vec{f}(t, \vec{y})\|$ .

## Numerical Solution of ODE IVPs

### Well-Posed versus Stable Solutions (2)

**Definition:** A solution of the ODE  $\vec{y}' = \vec{f}(t, \vec{y})$  is **stable** if for every  $\epsilon > 0$  there is a  $\delta > 0$  such that if  $\hat{\vec{y}}(t)$  satisfies the ODE and  $\|\hat{\vec{y}}(t_0) - \vec{y}(t_0)\| \leq \delta$  then  $\|\hat{\vec{y}}(t) - \vec{y}(t)\| \leq \epsilon$  for all  $t \geq t_0$ .

**Example:** The solution of the IVP  $y' = \lambda y$  with  $y(t_0) = y_0$  and  $\lambda$  a constant, is given by  $y(t) = y_0 e^{\lambda t}$ . So,

- If  $\lambda > 0$ , every solution is \_\_\_\_\_(1).
- If  $\lambda < 0$ , every solution is \_\_\_\_\_(2).

## Numerical Solution of ODE IVPs

### Fundamental Concepts: Euler's Method (1)

**Disclaimer:** Never use \_\_\_\_\_(1) actually to \_\_\_\_\_(2) an IVP. It is introduced here only to \_\_\_\_\_(3) basic concepts and definitions.

**Our canonical IVP:**

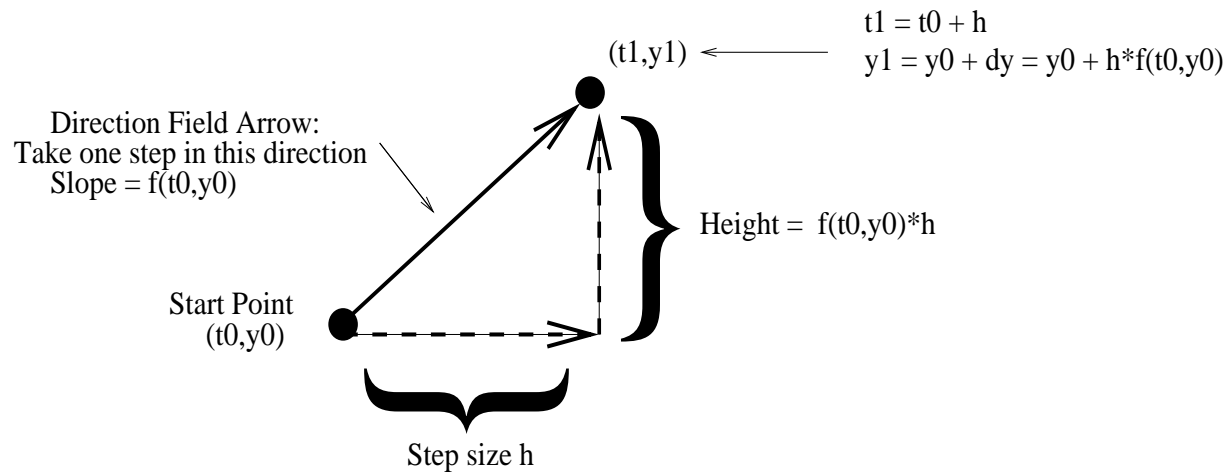
$$\frac{dy}{dt} = f(t, y), \quad y(t_0) = y_0$$

on some region  $D$  in the plane.

Assumption: All problems we will see are well-posed.

# Numerical Solution of ODE IVPs

## Fundamental Concepts: Euler's Method (2)



In general: Starting at  $(t_0, y_0)$ , we get to  $(t_{n+1}, y_{n+1})$  by using

$$t_{n+1} = t_n + h$$

$$y_{n+1} = y_n + h f(t_n, y_n)$$

**Note:**  $y_{n+1}$  represents the *numerical approximation* to  $y(t_{n+1})$ .

Euler's method is: \_\_\_\_\_(1) and \_\_\_\_\_(2).

# Numerical Solution of ODE IVPs

## Truncation Error

Truncation Error (TE) arises from the \_\_\_\_\_(1)  
of the true solution.

Usually, a \_\_\_\_\_(2) or \_\_\_\_\_(3) summation  
approximates an \_\_\_\_\_(4).

**Example:** Taylor expand  $y(t_{n+1})$  to get Euler's Method (a type of "Taylor Method"):

$$\underbrace{y(t_{n+1}) = y(t_n) + hy'(t_n)}_{\text{Euler's Method: Truncated Series}} + \underbrace{\frac{h^2}{2}y''(t_n) + \mathcal{O}(h^3)}_{\text{Local Truncation Error (LTE)}}$$



# Numerical Solution of ODE IVPs

## Order of a Method

Generally, Local Truncation Error (LTE) can be approximated by  $\alpha h^k$  in the sense that

$$\lim_{h \rightarrow 0} \left( \frac{\text{LTE}(h)}{h^k} \right) = \alpha$$

**In general:** The number  $k - 1$  is the **order** of the numerical method.

**Example:** Let  $y_1$  be the numerical approximation to  $y(t_1)$ . Then LTE in Euler's method is:

$$y(t_1) - y_1 = \frac{h^2}{2} y''(t_0) + \dots = \alpha h^2$$

Here,  $k =$  \_\_\_\_\_(1), so Euler's method is *order* \_\_\_\_\_(2).

**Terminology:**

An *order* \_\_\_\_\_(3) method implies:

The \_\_\_\_\_(4) **error behaves like** \_\_\_\_\_(5).

## Numerical Solution of ODE IVPs

### Summary: Types of Numerical Error

- Rounding error: From finite precision floating point arithmetic.  
(Example:  $\frac{1}{3} \approx 0.3333$ )
- Truncation error: From the method used. Two classes:
  - **Local (LTE)**: Error made in \_\_\_\_\_(1) of the numerical method.
  - **Global (GTE)**: Accumulated error. Error made after \_\_\_\_\_(2) of the numerical method.
- A numerical method is “order  $n$ ” if  $\text{LTE} = \mathcal{O}(h^{n+1})$ .

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: Introductory Example

Idea: If \_\_\_\_\_(1) do not cause the  
\_\_\_\_\_ (2) solution to diverge from the \_\_\_\_\_(3)  
solution, the numerical method is **stable**.

Example: Given the test IVP

$$y' = \lambda y, \quad y(0) = y_0$$

apply Euler's method with step size \_\_\_\_\_(4):

$$y_{n+1} = y_n + h\lambda y_n = (1 + h\lambda)y_n$$

which implies that

$$y_{n+1} = \underbrace{(1 + h\lambda)}_{\text{Amplification Factor}}^{n+1} y_0$$

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: Introductory Example (2)

**Example (continued):** The “Growth” or “Amplification Factor” =  $(1 + h\lambda)$ .

Therefore,

- If  $|1 + h\lambda| \leq 1$ , Euler’s method is \_\_\_\_\_(1).
- If  $|1 + h\lambda| > 1$ , Euler’s method is \_\_\_\_\_(2).

Requirements for Euler to be stable for this example:

- If  $\lambda$  **complex**:  $h\lambda$  must be inside a \_\_\_\_\_(3) disk in the complex plane centered at \_\_\_\_\_(4).
- If  $\lambda$  **real**:  $h\lambda$  must be in the interval \_\_\_\_\_(5).

**Note:** Choice of step-size  $h$  is crucial for stability.

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: General System

General System: \_\_\_\_\_(1)

Taylor Expand:  $\vec{y}(t + h) =$  \_\_\_\_\_(2)

Let  $t = t_k$  in Taylor expansion:

$$\vec{y}(t_{k+1}) = \vec{y}(t_k) + h\vec{f}(t, \vec{y}) + \mathcal{O}(h^2) \quad (1)$$

Euler's method:

$$\vec{y}_{k+1} = \vec{y}_k + h\vec{f}(t, \vec{y}) \quad (2)$$

Subtract equation (1) from equation (2):

$$\underbrace{\vec{y}_{k+1} - \vec{y}(t_{k+1})}_{\text{Global error } e_{k+1}} = (\vec{y}_k - \vec{y}(t_k)) + h \underbrace{\left( \vec{f}(t_k, \vec{y}_k) - \vec{f}(t_k, \vec{y}(t_k)) \right)}_{\text{Apply Mean Value Theorem}} - \mathcal{O}(h^2)$$

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: General System (2)

Applying the Mean Value Theorem:

$$\vec{f}(t_k, \vec{y}_k) - \vec{f}(t_k, \vec{y}(t_k)) = \mathbf{J}_f(t_k, \alpha \vec{y}_k + (1 - \alpha) \vec{y}(t_k)) (\vec{y}_k - \vec{y}(t_k))$$

where

- $\mathbf{J}_f =$  \_\_\_\_\_(1) matrix of  $\vec{f}$  w.r.t.  $\vec{y}$  and  $\alpha \in [0, 1]$ .

\_\_\_\_\_ (2) is expressed in general as:

$$\vec{e}_{k+1} = \underbrace{(\mathbf{I} + h\mathbf{J}_f)}_{\text{Amplification Factor}} \vec{e}_k + \text{LTE}_{k+1}$$

Requirement for \_\_\_\_\_(3):

$$\rho(\mathbf{I} + h\mathbf{J}) \leq 1$$

where  $\rho$  represents the \_\_\_\_\_(4) of a matrix.

## Numerical Solution of ODE IVPs

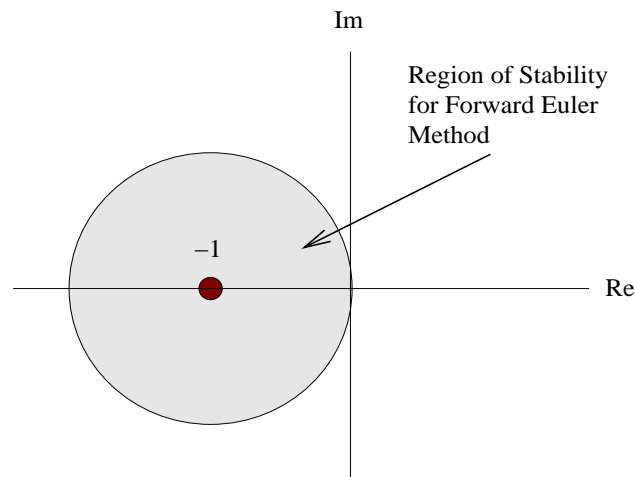
### Stability of a Numerical Method: General System (3)

- **Observation:** Stability requires that  $\rho(\mathbf{I} + h\mathbf{J}) \leq 1$ .
- **Question:** What does this stability restriction imply?
- **Answer:** All eigenvalues of  $h\mathbf{J}_f$  must lie inside a \_\_\_\_\_(1) disk in the \_\_\_\_\_(2) centered at \_\_\_\_\_(3).
- **Note:** If eigenvalues lie \_\_\_\_\_(4) the disk, the method will be \_\_\_\_\_(5).
- **Implication:** We must choose \_\_\_\_\_(6) so that all stability constraints are satisfied.

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: Euler Method Region of Stability

All eigenvalues of  $h\mathbf{J}_f$  must lie inside the disk.





## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: Euler Example

**Example:** Consider

$$y' = -10(t - 1)y, \quad y(0) = e^{-5}, \quad t \in [0, 2]$$

**Question:** When will Euler's Method be stable?

**Answer:** Notice that EM implies

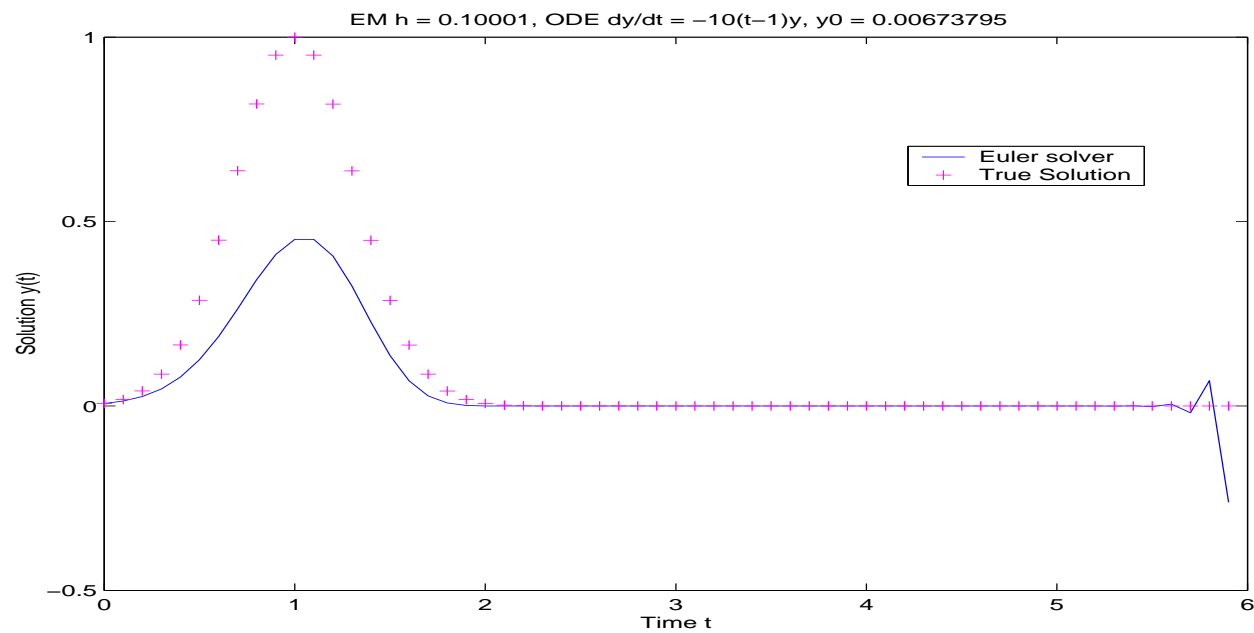
$$y_{k+1} = y_k + h(-10(t - 1)y_k) = \underbrace{(1 - 10h(t - 1))}_{\text{Amplification Factor}} y_k$$

Therefore,

- For \_\_\_\_\_(1), the method is unstable for any \_\_\_\_\_(2).
- For \_\_\_\_\_(3), the method will be stable if \_\_\_\_\_(4).

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: Euler Example Illustration



# Numerical Solution of ODE IVPs

## Implicit Methods

- **Recall:** Euler's method (EM) is \_\_\_\_\_(1) and \_\_\_\_\_(2). The limited region of stability for EM requires we choose \_\_\_\_\_(3) carefully.
- **Improvement:** Make the \_\_\_\_\_(4) \_\_\_\_\_(5).
- **How?** Use information at \_\_\_\_\_(6) as well as at \_\_\_\_\_(7).
- This makes the method \_\_\_\_\_(8).

# Numerical Solution of ODE IVPs

## Implicit Methods: Backward Euler

- Example of an Implicit Method: *Backward Euler Method (BE)*

$$\vec{y}_{k+1} = \vec{y}_k + h \underbrace{\vec{f}(t_{k+1}, \vec{y}_{k+1})}_{\text{Implicit}}$$

- **Question:** How can we solve for  $\vec{y}_{k+1}$  ?

- **Answer:**

- Use \_\_\_\_\_(1). This often requires calculating the \_\_\_\_\_(2) of the function  $\vec{f}(t, \vec{y})$ .

- Use \_\_\_\_\_(3) methods.

- **Note:** Both approaches require an initial guess, usually derived by taking one step of an explicit method.

# Numerical Solution of ODE IVPs

## Implicit Methods: Stability

- **Trade-off:** Implicit methods take \_\_\_\_\_(1) but are more \_\_\_\_\_(2).
- Apply BE to the test ODE  $y' = \lambda y$ :

$$y_{k+1} = y_k + h\lambda y_{k+1}$$
$$\text{Therefore } \Rightarrow y_k = \underbrace{\left(\frac{1}{1-h\lambda}\right)^k}_{\text{Amplification Factor}} y_0$$

- For BE to be stable we require

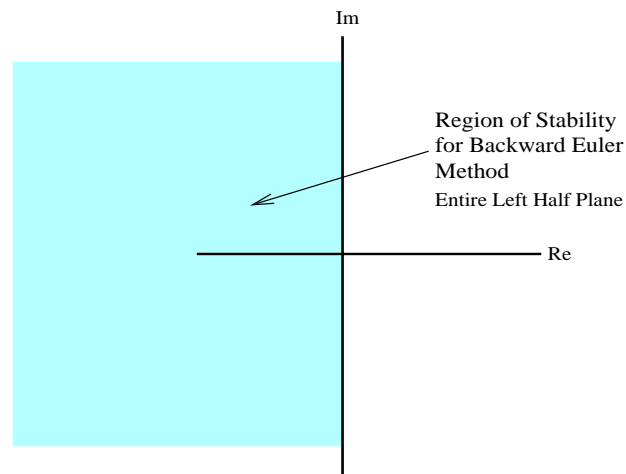
$$\left| \frac{1}{1-h\lambda} \right| < 1$$

- **Good news!** As long as  $\Re(\lambda) < 0$ , BE is stable for any step size  $h$ .

## Numerical Solution of ODE IVPs

### Stability of a Numerical Method: BE Stability Region Illustration

All eigenvalues of  $h\mathbf{J}_f$  must lie in the left half plane.



# Numerical Solution of ODE IVPs

## Implicit Methods: Stability Notes

### Notes:

- BE is \_\_\_\_\_(1) accurate (just like \_\_\_\_\_(2))
- For a \_\_\_\_\_(3), the stability requirement becomes  $\rho((\mathbf{I} - h\mathbf{J}_f)^{-1}) < 1$
- The stability region for BE is the entire \_\_\_\_\_(4) of the complex plane (as compared to the radius 1 disk of EM).
- Since \_\_\_\_\_(5) step size  $h$  keeps us within the stability region, the BE method is called \_\_\_\_\_(6).
- There exist higher order implicit methods, like the Trapezoid Method (order = \_\_\_\_\_(7)).
- Not all implicit methods are \_\_\_\_\_(8).

# Numerical Solution of ODE IVPs

## Stiffness in ODEs

**Question:** What is “stiffness”?

**Answer:**

- Physically: A process whose components have \_\_\_\_\_(1) time scales. Also, a process whose time scale is \_\_\_\_\_(2) compared to the time interval over which it is being observed.
- Mathematically: A well-posed ODE  $\vec{y}' = \vec{f}(t, \vec{y})$  is “stiff” if its Jacobian  $\mathbf{J}_f$  has \_\_\_\_\_(3) that differ greatly in magnitude.
- Practically: An ODE is stiff if an explicit method (like EM) is \_\_\_\_\_(4) because stability requirements force the step size  $h$  to be extremely small.



## Numerical Solution of ODE IVPs

### Stiffness in ODEs: Example 1

**Example 1:** For test ODE  $y' = \lambda y$  with  $t \in [a, b]$ , the problem is

\_\_\_\_\_ (1) if \_\_\_\_\_ (2).

Recall: For stability in EM, we require  $|1 - h\lambda| < 1$ . If  $\lambda$  is real and  $\lambda \ll -1$ , this forces \_\_\_\_\_ (3).

For a **system**: A system of ODEs with  $t \in [a, b]$  is \_\_\_\_\_ (4) when

\_\_\_\_\_ (5) where  $\lambda_j$  are the

\_\_\_\_\_ (6) of Jacobian  $\mathbf{J}(t, \vec{y}(t))$ .

## Numerical Solution of ODE IVPs

### Stiffness in ODEs: Example 2

Example 2: Consider

$$y' = -\alpha(y - \sin t) + \cos t, \quad y(0) = 1, \quad t \in [0, 1]$$

Let  $\alpha = 1000$ . Then the Jacobian is \_\_\_\_\_(1). So eigenvalue  $\lambda =$  \_\_\_\_\_(2). Therefore,

$$(1 - 0)\Re(\lambda) = (-\alpha) = -1000 \ll -1$$

This is \_\_\_\_\_(3) on  $t \in [0, 1]$ .

Note: on another interval,  $t \in [0, 0.002]$  we have

$$(0.002 - 0)\Re(\lambda) = (0.002)(-1000) = -2$$

so this ODE is \_\_\_\_\_(4) on this interval.

## Numerical Solution of ODE IVPs

### Stiffness: EM vs BE

Consider

$$y' = -100y + 100t + 101, \quad y(0) = 1$$

Let  $h = 0.1$ . Computed output with perturbed initial data:

Time $t$	0.00	0.10	0.20	0.30	0.40
Exact Soln	<b>1.00</b>	<b>1.10</b>	<b>1.20</b>	<b>1.30</b>	<b>1.40</b>
EM	0.99	1.19	0.39	8.59	-64.21
EM	1.01	1.01	2.01	-5.99	67.01
BE	0.00	1.01	1.19	1.30	1.40
BE	2.00	1.19	1.21	1.30	1.40

## Numerical Solution of ODE IVPs

### Stiffness: Comments on EM vs BE

Note:

- EM \_\_\_\_\_(1) with only \_\_\_\_\_(2) perturbations.
- BE is \_\_\_\_\_(3) even with \_\_\_\_\_(4) perturbations.

## Numerical Solution of ODE IVPs

### Stiffness: Summary

- A particular ODE may be \_\_\_\_\_(1) or \_\_\_\_\_(2).
- A numerical ODE solving method can be \_\_\_\_\_(3) or \_\_\_\_\_(4) for a particular problem.
- The stability of the numerical method often depends on \_\_\_\_\_(5).
- \_\_\_\_\_(6) methods should (almost) always be used to solve stiff ODEs.

# Numerical Solution of ODE IVPs

## Other IVP Solvers

Other classes of numerical IVP solvers include (but are not limited to):

- Higher Order Taylor methods (seldom used)
  - Can give \_\_\_\_\_(1) accuracy.
  - They require the computation of the \_\_\_\_\_(2) of  $f(t, y)$ .
- Runge-Kutta methods (very popular)
  - Can give \_\_\_\_\_(3) accuracy.
  - Do not need \_\_\_\_\_(4) of  $f(t, y)$ .
  - Can be \_\_\_\_\_(5) or \_\_\_\_\_(6).
  - These are \_\_\_\_\_(7)-step methods.
  - Methods include: Midpoint, Modified Euler, Heun, 4th Order Runge-Kutta (RK4)

# Numerical Solution of ODE IVPs

## Other IVP Solvers (cont)

- Multi-step methods

- Can give \_\_\_\_\_(1) accuracy.
- Can be \_\_\_\_\_(2) or \_\_\_\_\_(3).
- Starting values must be calculated with a \_\_\_\_\_(4) method (e.g., RK4)
- Methods include: Adams-Bashforth, Adams-Moulton, Milne, Simpson

- Extrapolation methods

- These take solutions generated by lower order methods, and increase \_\_\_\_\_(5) by \_\_\_\_\_(6).
- Variations of these methods presented in [SB93], [Ste73], and [Gra65]. Also see the discussion and reference list in [Asa95, p.642].

## Numerical Solution of ODE IVPs

### Systems and Higher Order IVPs

- All methods and theories presented can be \_\_\_\_\_(1) to apply to systems.
- Many higher order IVPs can be \_\_\_\_\_(2) to 1st order systems of IVPs. Then all methods and theories apply here, too.

**Example:** Suppose we have the second order equation describing a linear spring,

$$y'' = -ky, \quad y(0) = y_0, \quad y'(0) = v_0$$

Convert this to a  $2 \times 2$  first order system of equations:

$$\begin{aligned} y' &= v \\ v' &= \text{_____}(3) \end{aligned}$$



## Numerical Solution of ODE IVPs

### Solving IVPs with Packaged Software

To solve a system of ODE IVPs  $\vec{y}' = \vec{f}(t, \vec{y})$  with  $y(t_0) = y_0$  with a packaged routine typically requires \_\_\_\_\_(1) to supply the following:

- The name of the routine that computes  $\vec{f}(t, \vec{y})$ .
- \_\_\_\_\_(2) and \_\_\_\_\_(3) values for times  $t$ .
- Initial value  $y_0$ .
- ... and for some solvers, sometimes ...
- The number of equations in the system.
- \_\_\_\_\_(4) and/or \_\_\_\_\_(5) error tolerances.
- ... and sometimes for a stiff ODE ...
- The routine that computes the Jacobian  $\mathbf{J}_f$  of function  $\vec{f}$ .