

Stability Analysis: multistep methods (I)

single ODE $\frac{dy}{dt} = f(t, y), \quad a \leq t \leq b, \quad y(a) = \alpha.$

Consider multistep method, $w_0 = \alpha, w_1 = \alpha_1, \dots, w_{m-1} = \alpha_{m-1}$,

- ▶ for $j = m - 1, m, m + 1, \dots$

$$\begin{aligned}w_{j+1} &= a_{m-1} w_j + a_{m-2} w_{j-1} + \cdots + a_0 w_{j+1-m} \\&\quad + h F(t_j, w_{j+1}, w_j, \dots, w_{j+1-m}), \quad x_j = a + j h.\end{aligned}$$

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- ▶ local truncation error

$$\begin{aligned}\text{LTE } \tau_{j+1}(h) &\stackrel{\text{def}}{=} \frac{y(t_{j+1}) - (a_{m-1} y(t_j) + a_{m-2} y(t_{j-1}) + \cdots + a_0 y(t_{j+1-m}))}{h} \\&\quad - F(t_j, y(t_{j+1}), y(t_j), \dots, y(t_{j+1-m})).\end{aligned}$$

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Assumptions on F

- ▶ If $f \equiv 0$, then $F \equiv 0$.

$$\begin{aligned} &|F(t_j, u_{j+1}, u_j, \dots, u_{j+1-m}) - F(t_j, \hat{u}_{j+1}, \hat{u}_j, \dots, \hat{u}_{j+1-m})| \\ &\leq L(|u_{j+1} - \hat{u}_{j+1}| + \dots + |u_{j+1-m} - \hat{u}_{j+1-m}|) \end{aligned}$$

Stability Analysis: multistep methods (II)

- ▶ *Definition: consistency*

$$\lim_{h \rightarrow 0} \max_{m \leq j \leq N} |\tau_j(h)| = 0, \quad \lim_{h \rightarrow 0} \max_{0 \leq j \leq m-1} |y(t_j) - \alpha_j| = 0.$$

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Stability is a much bigger issue

Stability Analysis: multistep methods (III)

single ODE

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Multistep method with $w_0 = \alpha, w_1 = \alpha_1, \dots, w_{m-1} = \alpha_{m-1}$,

- ▶ for $j = m - 1, m, m + 1, \dots$

$$\begin{aligned} w_{j+1} &= a_{m-1} w_j + a_{m-2} w_{j-1} + \dots + a_0 w_{j+1-m} \\ &\quad + h F(t_j, w_{j+1}, w_j, \dots, w_{j+1-m}), \\ &= a_{m-1} w_j + a_{m-2} w_{j-1} + \dots + a_0 w_{j+1-m}. \quad (F \equiv 0) \end{aligned}$$

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- ▶ Assume $\alpha = \alpha_1 \dots = \alpha_{m-1}$,

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- ▶ Assume $\alpha = \alpha_1 = \dots = \alpha_{m-1}$,

Minimum requirements

- ▶ $w_{j+1} \equiv \alpha$ for all j .
- ▶ w_{j+1} remains close to α in finite precision.

Finite recurrence relations (I)

Given $w_0 = \alpha_0, w_1 = \alpha_1, \dots, w_{m-1} = \alpha_{m-1}$,

- ▶ for $j = m - 1, m, m + 1, \dots$

$$w_{j+1} = a_{m-1} w_j + a_{m-2} w_{j-1} + \dots + a_0 w_{j+1-m}.$$

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$$w_{j+1} = a_{m-1} w_j + a_{m-2} w_{j-1} + \dots + a_0 w_{j+1-m}.$$

- ▶ To solve for w_j for all j , **assume**

$$\frac{w_{k+1}}{w_k} = \lambda \quad \text{for all } k.$$

- ▶ Recurrence becomes

$$P(\lambda) = 0, \quad P(\lambda) \stackrel{\text{def}}{=} \lambda^m - (a_{m-1} \lambda^{m-1} + a_{m-2} \lambda^{m-2} + \dots + a_0).$$

- ▶ thus λ must be a root of $P(\lambda)$.
- ▶ $\lambda = 1$ should be a root of $P(\lambda)$.

Finite recurrence relations (II)

- ▶ Recurrence relation

$$w_{j+1} = a_{m-1} w_j + a_{m-2} w_{j-1} + \cdots + a_0 w_{j+1-m}.$$

- ▶ characteristic polynomial

$$\mathbf{P}(\lambda) \stackrel{\text{def}}{=} \lambda^m - (a_{m-1} \lambda^{m-1} + a_{m-2} \lambda^{m-2} + \cdots + a_0).$$

- ▶ If $\mathbf{P}(\lambda)$ has m **distinct** roots $\lambda_1, \dots, \lambda_m$, then

$$w_j = c_1 \lambda_1^j + c_2 \lambda_2^j + \cdots + c_m \lambda_m^j, \quad j = 0, 1, \dots, m-1, m, \dots$$

for constants c_1, c_2, \dots, c_m determined by the equations for
 $0 \leq j \leq m-1$.

Finite recurrence relations (III)

- ▶ Example recurrence relation

$$w_{j+1} = 3w_j - 2w_{j-1}. \quad (m = 2.)$$

- ▶ **characteristic polynomial**

$$\mathbf{P}(\lambda) = \lambda^2 - 3\lambda^1 + 2 = (\lambda - 1)(\lambda - 2).$$

- ▶ Roots of $\mathbf{P}(\lambda)$ are 1 and 2.
- ▶ recurrence solution

$$w_j = c_1 + c_2 2^j, \quad j = 0, 1, 2, 3, \dots$$

where

$$w_0 = c_1 + c_2, \quad w_1 = c_1 + 2c_2, \quad \text{or}$$

$$c_1 = 2w_0 - w_1, \quad c_2 = w_1 - w_0.$$

Finite recurrence relations (IV)

- ▶ Example recurrence relation

$$w_{j+1} = 2w_j - 1 w_{j-1}. \quad (m = 2.)$$

- ▶ **characteristic polynomial**

$$\mathbf{P}(\lambda) = \lambda^2 - 2\lambda^1 + 1 = (\lambda - 1)^2.$$

- ▶ Roots of $\mathbf{P}(\lambda)$ are 1 and 1.
- ▶ recurrence solution

$$w_j = c_1 + j c_2, \quad j = 0, 1, 2, 3, \dots$$

where

$$w_0 = c_1, \quad w_1 = w_1 - w_0.$$

Root conditions

Multistep method with $w_0 = \alpha, w_1 = \alpha_1, \dots, w_{m-1} = \alpha_{m-1}$,

- ▶ for $j = m - 1, m, m + 1, \dots$

$$\begin{aligned}w_{j+1} &= a_{m-1} w_j + a_{m-2} w_{j-1} + \cdots + a_0 w_{j+1-m} \\&\quad + h F(t_j, w_{j+1}, w_j, \dots, w_{j+1-m}),\end{aligned}$$

$$\mathbf{P}(\lambda) = \lambda^m - (a_{m-1} \lambda^{m-1} + a_{m-2} \lambda^{m-2} + \cdots + a_0).$$

Root conditions

Multistep method with $w_0 = \alpha, w_1 = \alpha_1, \dots, w_{m-1} = \alpha_{m-1}$,

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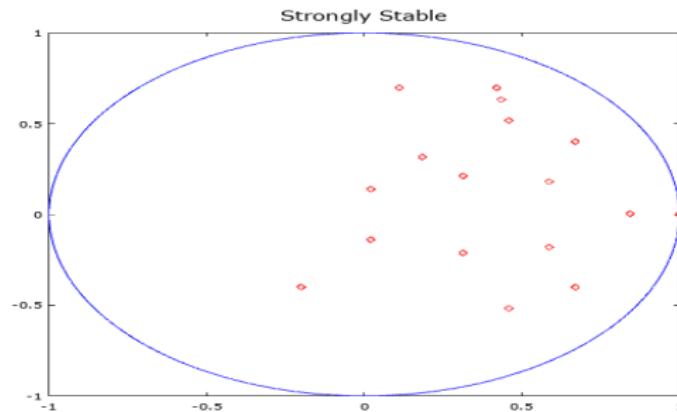
root condition: every root λ_i of $\mathbf{P}(\lambda)$ must satisfy $|\lambda_i| \leq 1$

Assume multistep method satisfies root condition.

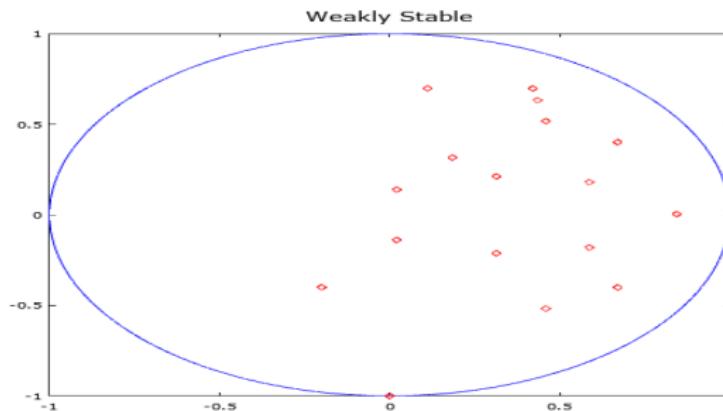
- ▶ **strongly stable:** $\lambda = 1$ is only root of $\mathbf{P}(\lambda)$ with magnitude 1.
- ▶ **weakly stable:** $\mathbf{P}(\lambda)$ has more than 1 distinct root with magnitude 1.

Otherwise method is **unstable**.

► **strongly stable:**



► **weakly stable:**



single ODE

$$\frac{dy}{dt} = f(t, y), \quad a \leq t \leq b, \quad y(a) = \alpha.$$

Theorem: Assume multistep method with $w_0 = \alpha$,

$$w_1 = \alpha_1, \dots, w_{m-1} = \alpha_{m-1},$$

- ▶ for $j = m - 1, m, m + 1, \dots$

$$\begin{aligned} w_{j+1} &= a_{m-1} w_j + a_{m-2} w_{j-1} + \cdots + a_0 w_{j+1-m} \\ &\quad + h F(t_j, w_{j+1}, w_j, \dots, w_{j+1-m}). \end{aligned}$$

single ODE

$$\frac{dy}{dt} = f(t, y), \quad a \leq t \leq b, \quad y(a) = \alpha.$$

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Assume the method is **consistent**, then

- ▶ The method is stable \iff it satisfies root condition
 \iff it is convergent.

example fourth-order Adams-Bashforth method

- ▶ 4-step and fourth-order

$$w_{j+1} = w_j + h F(t_j, w_j, w_{j-1}, w_{j-2}, w_{j-3})$$

where $F(t_j, w_j, w_{j-1}, w_{j-2}, w_{j-3}) =$

$$\frac{h}{24} (55f(t_j, w_j) - 59f(t_{j-1}, w_{j-1}) + 37f(t_{j-2}, w_{j-2}) - 9f(t_{j-3}, w_{j-3})).$$

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Solution: $m = 4$,

$$\mathbf{P}(\lambda) = \lambda^4 - \lambda^3 = \lambda^3(\lambda - 1).$$

Roots of $\mathbf{P}(\lambda)$ are $0, 0, 0, 1$, satisfying root condition:
strongly stable.

example fourth-order Milne's method

- ▶ 4-step and fourth-order

$$w_{j+1} = w_{j-3} + h \hat{F}(t_j, w_j, w_{j-1}, w_{j-2}, w_{j-3})$$

where $\hat{F}(t_j, w_j, w_{j-1}, w_{j-2}, w_{j-3})$
 $= \frac{4h}{3} (2f(t_j, w_j) - f(t_{j-1}, w_{j-1}) + 2f(t_{j-2}, w_{j-2}))$.

example fourth-order Milne's method

- ▶ 4-step and fourth-order

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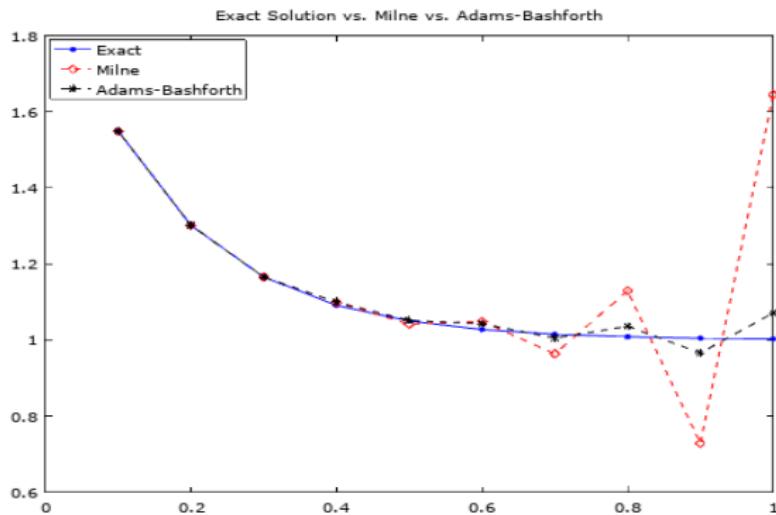
$$\mathbf{P}(\lambda) = \lambda^4 - 1 = (\lambda - 1)(\lambda + 1)(\lambda - \sqrt{-1})(\lambda + \sqrt{-1}).$$

Roots of $\mathbf{P}(\lambda)$ are $\pm 1, \pm \sqrt{-1}$, satisfying root condition:
weakly stable as all roots have magnitude 1.

Example: Adams-Bashforth vs. Milne

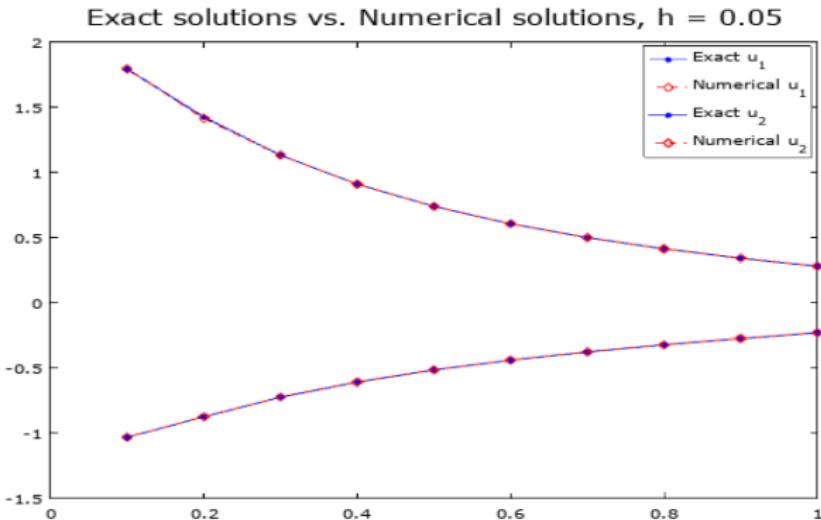
$$\frac{dy}{dt} = -6y + 6, \quad 0 \leq t \leq 1, \quad y(0) = 2,$$

exact solution $y(t) = 1 + e^{-6t}$, $h = 0.1$.



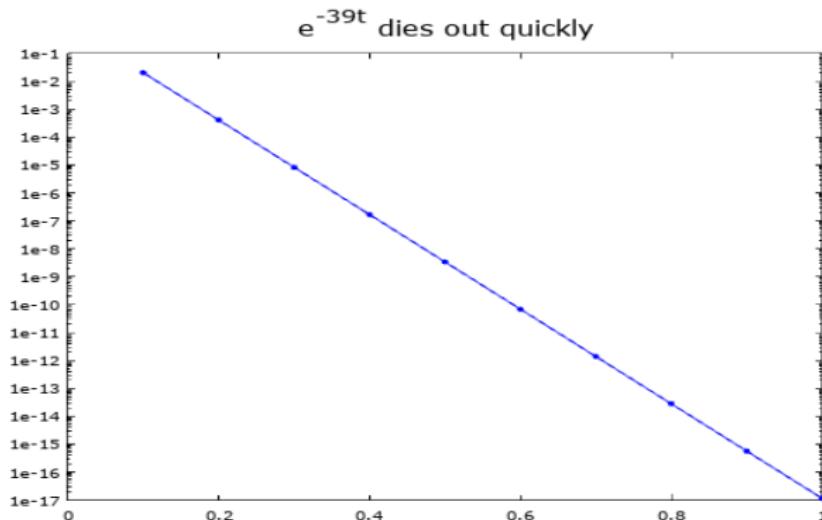
Example: Stiff ODEs (I), $h = 0.05$

$$\frac{d}{dt} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 9u_1 + 24u_2 + 5\cos t - \frac{1}{3}\sin t, \\ -24u_1 - 51u_2 - 9\cos t + \frac{1}{3}\sin t \end{pmatrix}, \quad \begin{pmatrix} u_1(0) \\ u_2(0) \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 4 \\ 2 \end{pmatrix},$$
$$\begin{pmatrix} u_1(t) \\ u_2(t) \end{pmatrix} = \begin{pmatrix} 2e^{-3t} - e^{-39t} + \frac{1}{3}\cos t \\ -e^{-3t} + 2e^{-39t} - \frac{1}{3}\cos t \end{pmatrix}, \quad \boxed{\text{exact solution, zombie term } e^{-39t}}$$



Transient e^{-39t} : little effect on solution, big effect on h

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Example: Stiff ODEs (III), $h = 0.1$

$$\frac{d}{dt} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 9u_1 + 24u_2 + 5\cos t - \frac{1}{3}\sin t, \\ -24u_1 - 51u_2 - 9\cos t + \frac{1}{3}\sin t \end{pmatrix}, \quad \begin{pmatrix} u_1(0) \\ u_2(0) \end{pmatrix} = \frac{1}{3} \begin{pmatrix} 4 \\ 2 \end{pmatrix},$$
$$\begin{pmatrix} u_1(t) \\ u_2(t) \end{pmatrix} = \begin{pmatrix} 2e^{-3t} - e^{-39t} + \frac{1}{3}\cos t \\ -e^{-3t} + 2e^{-39t} - \frac{1}{3}\cos t \end{pmatrix}, \quad \boxed{\text{exact solution, zombie term } e^{-39t}}$$

t_j	numerical u_1	numerical u_2
0.1	-2.6452	7.8445
0.2	-18.452	38.876
0.3	-87.472	176.48
0.4	-934.07	789.35
0.5	-1760	3520
0.6	-7848.6	15698
0.7	-34990	69980
0.8	$-1.5598e + 05$	$3.1196e + 05$
0.9	$-6.9533e + 05$	$1.3907e + 06$
1.0	$-3.0997e + 06$	$6.1994e + 06$

Test equations for stiff ODEs

$$\frac{dy}{dt} = \lambda y, \quad y(0) = \alpha, \quad \text{for } \lambda < 0.$$

- ▶ test equation has transient solution $y(t) = \alpha e^{\lambda t}$ that decays to 0 as $t \rightarrow 0$.
- ▶ test equation has steady-state solution $y(t) = 0$.

Desired of numerical methods:

- ▶ **convergence:**

$$\lim_{h \rightarrow 0} |y(t_j) - w_j| = 0.$$

- ▶ **numerical stability:** small error in α leads to small error in w_j .

Euler's Method (I)

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- ▶ Euler's method: $w_0 = \alpha$,

$$w_{j+1} = w_j + h \lambda w_j = (1 + \lambda h) w_j, \quad j = 0, 1, \dots,$$

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$$\begin{aligned} w_j &= (1 + \lambda h)^j w_0 = (1 + \lambda h)^j \alpha, \\ |y(t_j) - w_j| &= \left| e^{\lambda h j} - (1 + \lambda h)^j \right| |\alpha|. \end{aligned}$$

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- ▶ for convergence, need

$$|1 + \lambda h| < 1, \quad \text{or} \quad -2 < \lambda h < 0.$$

Euler's Method (II)

- ▶ Euler's method: $w_0 = \alpha$, and for $j = 0, 1, \dots$,

$$w_{j+1} = (1 + \lambda h) w_j = (1 + \lambda h)^{j+1} \alpha.$$

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- ▶ Now assume a round-off error of δ in w_0 :

$$\hat{w}_0 = \alpha + \delta.$$

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- ▶ Euler's method numerically produces, for $j = 0, 1, \dots$,

$$\hat{w}_{j+1} = (1 + \lambda h) \hat{w}_j = (1 + \lambda h)^{j+1} (\alpha + \delta).$$

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Multistep Methods (I)

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- ▶ general multistep method, for $j = m - 1, m, \dots$,

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- ▶ **region R of absolute stability**

$$\mathbf{R} \stackrel{\text{def}}{=} \{ \lambda h \in \mathbb{C} \mid |\beta_k| < 1 \text{ for all zeros } \beta_k \text{ of } \mathbf{Q}(z, \lambda h). \}$$

region R of absolute stability: Euler's method

test equation $\frac{dy}{dt} = \lambda y, \quad y(0) = \alpha, \quad \text{for } \lambda < 0.$

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$$w_{j+1} = w_j + h f(t_j, w_j) = (1 + \lambda h) w_j.$$

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$$\mathbf{Q}(z, \lambda h) = z - (1 + \lambda h) = 0, \quad \text{with root } \beta_1 = 1 + \lambda h.$$

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$$\mathbf{R} = \{ \lambda h \in \mathbb{C} \mid |1 + \lambda h| < 1. \}$$

region R of absolute stability: Implicit Trapezoid

test equation $\frac{dy}{dt} = \lambda y, \quad y(0) = \alpha, \quad \text{for } \lambda < 0.$

- ▶ Implicit Trapezoid

$$w_{j+1} = w_j + \frac{h}{2} (f(t_{j+1}, w_{j+1}) + f(t_j, w_j)) = w_j + \frac{\lambda h}{2} (w_{j+1} + w_j).$$

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- ▶ Definition: Implicit Trapezoid is **A-stable**.

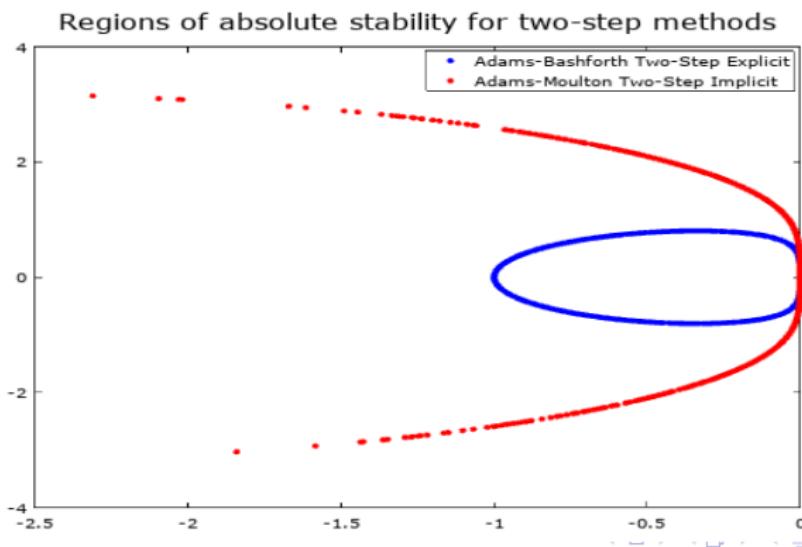
regions of absolute stability:

- ▶ Adams-Bashforth Two-Step Explicit Method

$$w_{j+1} = w_j + \frac{h}{2} (3f(t_j, w_j) - f(t_{j-1}, w_{j-1})).$$

- ▶ Adams-Moulton Two-Step Implicit Method

$$w_{j+1} = w_j + \frac{h}{12} (5f(t_{j+1}, w_{j+1}) + 8f(t_j, w_j) - f(t_{j-1}, w_{j-1})).$$



Implicit Trapezoid with Newton Iteration

► Implicit Trapezoid

$$w_{j+1} = w_j + \frac{h}{2} (f(t_{j+1}, w_{j+1}) + f(t_j, w_j)).$$

Implicit Trapezoid with Newton Iteration

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$$\mathbf{F}(w) \stackrel{\text{def}}{=} w - w_j - \frac{h}{2} (f(t_{j+1}, w) + f(t_j, w_j)) = 0.$$

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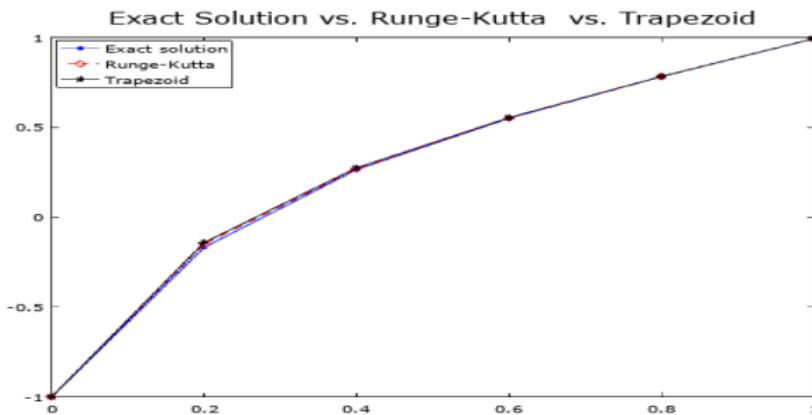
- ▶ Newton Iteration: $w_{j+1}^{(0)} = w_j$ (no predictor); for $\ell = 0, 1, \dots$

$$\begin{aligned} w_{j+1}^{(\ell+1)} &= w_{j+1}^{(\ell)} - \frac{\mathbf{F}\left(w_{j+1}^{(\ell)}\right)}{\mathbf{F}'\left(w_{j+1}^{(\ell)}\right)} \\ &= w_{j+1}^{(\ell)} - \frac{w_{j+1}^{(\ell)} - w_j - \frac{h}{2} \left(f(t_{j+1}, w_{j+1}^{(\ell)}) + f(t_j, w_j)\right)}{1 - \frac{h}{2} \frac{\partial f}{\partial y}(t_{j+1}, w_{j+1}^{(\ell)})} \end{aligned}$$

Implicit Trapezoid example (I)

$$\begin{aligned}\frac{dy}{dt} &= 5e^{5t}(y - t)^2 + 1, \quad 0 \leq t \leq 1, \quad y(0) = -1, \\ y(t) &= t - e^{-5t} \quad \text{exact solution}\end{aligned}$$

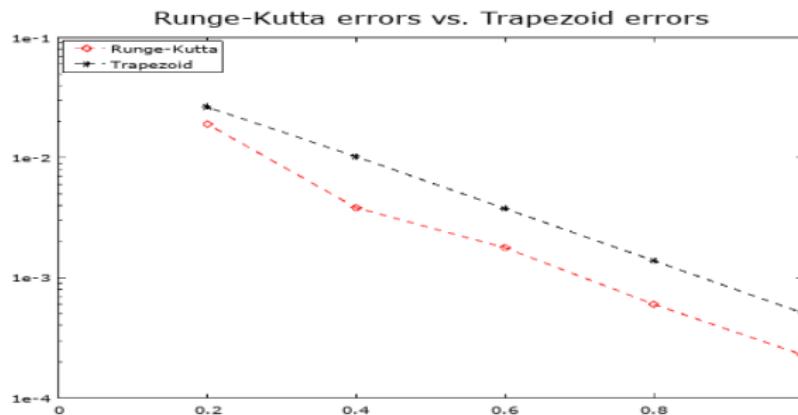
- ▶ Implicit Trapezoid vs. 4th order Runge-Kutta , $h = 0.2$



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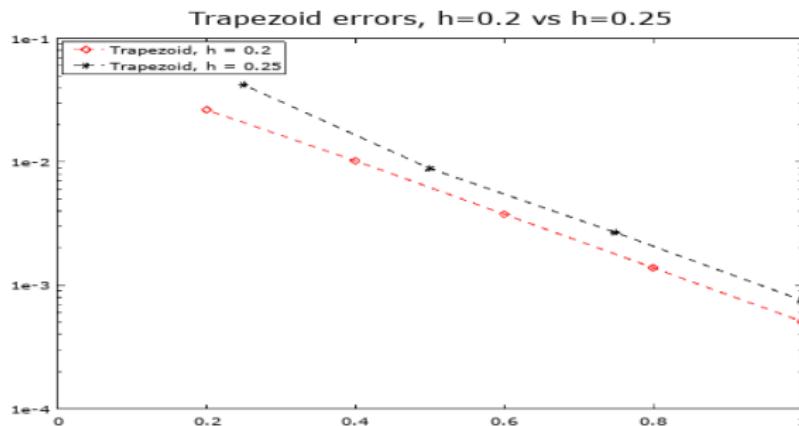
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Implicit Trapezoid example (II)

$$\begin{aligned}\frac{dy}{dt} &= 5e^{5t}(y - t)^2 + 1, \quad 0 \leq t \leq 1, \quad y(0) = -1, \\ y(t) &= t - e^{-5t} \quad \text{exact solution}\end{aligned}$$

- ▶ Implicit Trapezoid, $h = 0.25$



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t_j	4 th order Runge Kutta
0.0	-1.0
0.25	0.4014315
0.5	3.4374753
0.75	$1.44639e + 23$
1.0	Inf

Linear Equations: example of ellipse $\frac{x^2}{a^2} + \frac{y^2}{b^2} = 1$.

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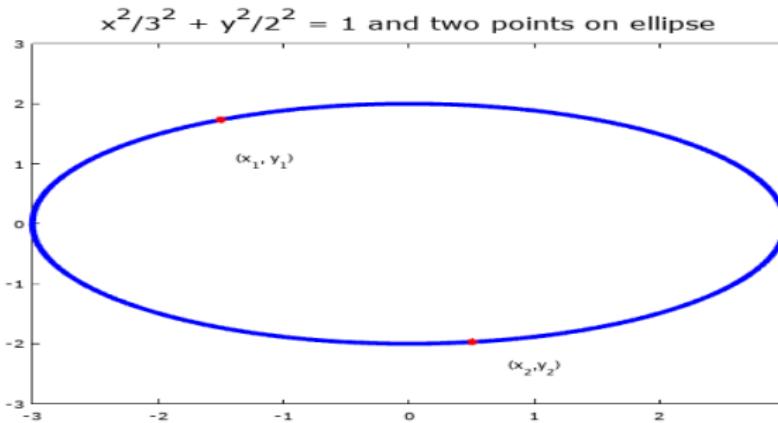
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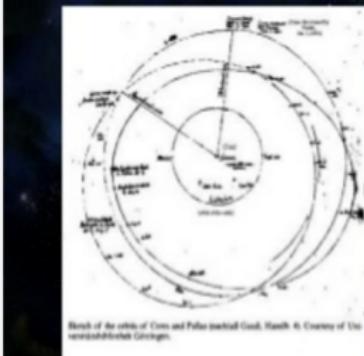
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Carl Gauss: World's first numerical analyst

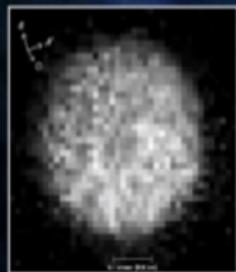


(Left) Carl Friedrich Gauss, considered one of the three greatest mathematicians of all time (along with Archimedes and Sir Isaac Newton).
(Right) Gauss at 24, when he computed the orbit of Ceres.



(Left) Gauss' sketch of the orbit of Ceres.

(Right) Image of Ceres from the Hubble telescope.



Solving Linear Equations

- ▶ example: solving for x_1, x_2, x_3, x_4 in system of linear equations

$$E_1 : \quad x_1 \quad + \quad x_2 \quad + \quad 3x_4 \quad = \quad 4,$$

$$E_2 : \quad 2x_1 \quad + \quad x_2 \quad - \quad x_3 \quad + \quad x_4 \quad = \quad 1,$$

$$E_3 : \quad 3x_1 \quad - \quad x_2 \quad - \quad x_3 \quad + \quad 2x_4 \quad = \quad -3,$$

$$E_4 : \quad -x_1 \quad + \quad 2x_2 \quad + \quad 3x_3 \quad - \quad x_4 \quad = \quad 4.$$

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Solution techniques (elementary operations)

- ▶ Multiply equation E_i by any constant $\lambda \neq 0$, with the resulting equation used in place of E_i .
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- ▶ Equations E_i and E_j can be transposed in order.
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Solving Linear Equations: example (I)

- ▶ example: solving for x_1, x_2, x_3, x_4 in system of linear equations

$$E_1 : \quad x_1 + x_2 + 3x_4 = 4,$$

$$E_2 : \quad 2x_1 + x_2 - x_3 + x_4 = 1,$$

$$E_3 : \quad 3x_1 - x_2 - x_3 + 2x_4 = -3,$$

$$E_4 : \quad -x_1 + 2x_2 + 3x_3 - x_4 = 4.$$

- ▶ eliminate x_1 from E_2 , $(E_2 - 2E_1) \rightarrow (E_2)$:

$$(2x_1 + x_2 - x_3 + x_4) - 2(x_1 + x_2 + 3x_4) = 1 - 2 \times 4.$$

- ▶ new system of equations, same solution in x_1, x_2, x_3, x_4

$$E_1 : \quad x_1 + x_2 + 3x_4 = 4,$$

$$E_2 : \quad -x_2 - x_3 - 5x_4 = -7,$$

$$E_3 : \quad 3x_1 - x_2 - x_3 + 2x_4 = -3,$$

$$E_4 : \quad -x_1 + 2x_2 + 3x_3 - x_4 = 4.$$

Solving Linear Equations: example (II)



$$\begin{aligned}E_1 : \quad &x_1 + x_2 + 3x_4 = 4, \\E_2 : \quad &-x_2 - x_3 - 5x_4 = -7, \\E_3 : \quad &3x_1 - x_2 - x_3 + 2x_4 = -3, \\E_4 : \quad &-x_1 + 2x_2 + 3x_3 - x_4 = 4.\end{aligned}$$

- eliminate x_1 from E_3, E_4 :

$$(E_3 - 3E_1) \rightarrow (E_3), \quad (E_4 - (-1)E_1) \rightarrow (E_4).$$

- new system of equations, same solution in x_1, x_2, x_3, x_4

$$\begin{aligned}E_1 : \quad &x_1 + x_2 + 3x_4 = 4, \\E_2 : \quad &-x_2 - x_3 - 5x_4 = -7, \\E_3 : \quad &-4x_2 - x_3 - 7x_4 = -15, \\E_4 : \quad &+ 3x_2 + 3x_3 + 2x_4 = 8.\end{aligned}$$

- equations E_2, E_3, E_4 no longer contain x_1 .

Solving Linear Equations: example (III)



$$\begin{aligned}E_1 : \quad & x_1 + x_2 + 3x_4 = 4, \\E_2 : \quad & -x_2 - x_3 - 5x_4 = -7, \\E_3 : \quad & -4x_2 - x_3 - 7x_4 = -15, \\E_4 : \quad & + 3x_2 + 3x_3 + 2x_4 = 8.\end{aligned}$$

- ▶ Use E_2 to eliminate x_2 from E_3 and E_4 by performing

$$(E_3 - 4E_2) \rightarrow (E_3), \quad (E_4 - (-3)E_2) \rightarrow (E_4).$$

- ▶ new system of equations,

$$\begin{aligned}E_1 : \quad & x_1 + x_2 + 3x_4 = 4, \\E_2 : \quad & -x_2 - x_3 - 5x_4 = -7, \\E_3 : \quad & 3x_3 + 13x_4 = 13, \\E_4 : \quad & - 13x_4 = -13.\end{aligned}$$

Solving Linear Equations: example (IV), backward substitution

$$\begin{array}{l} \text{E}_1 : x_1 + x_2 + 3x_4 = 4 \\ \text{E}_2 : -x_2 - x_3 - 5x_4 = -7 \\ \text{E}_3 : 3x_3 + 13x_4 = 13 \\ \text{E}_4 : -13x_4 = -13 \end{array} \quad \left| \begin{array}{l} x_1 = 4 - x_2 - 3x_4 = -1, \\ x_2 = \frac{-7 + x_3 + 5x_4}{-1} = 2, \\ x_3 = \frac{13 - 13x_4}{3} = 0, \\ x_4 = \frac{-13}{-13} = 1. \end{array} \right.$$

Solving Linear Equations with pivoting (I)

- solving for x_1, x_2, x_3, x_4

$$E_1 : \quad x_1 - x_2 + 2x_3 - x_4 = -8,$$

$$E_2 : \quad 2x_1 - 2x_2 + 3x_3 - 3x_4 = -20,$$

$$E_3 : \quad x_1 + x_2 + x_3 = -2,$$

$$E_4 : \quad x_1 - x_2 + 4x_3 + 3x_4 = 4.$$

- eliminate x_1 from E_2, E_3 and E_4 :

$$(E_2 - 2E_1) \rightarrow (E_2), \quad (E_3 - E_1) \rightarrow (E_3), \quad (E_4 - E_1) \rightarrow (E_4).$$

- new system of equations,

$$E_1 : \quad x_1 - x_2 + 2x_3 - x_4 = -8,$$

$$E_2 : \quad -x_3 - x_4 = -4,$$

$$E_3 : \quad 2x_2 - x_3 + x_4 = 6,$$

$$E_4 : \quad 2x_3 + 4x_4 = 12.$$

- Can NOT use E_2 to eliminate x_2 from E_3 and E_4

Solving Linear Equations with pivoting (II)



$$\begin{aligned}E_1 : \quad x_1 &- x_2 + 2x_3 - x_4 = -8, \\E_2 : \quad &\quad - x_3 - x_4 = -4, \\E_3 : \quad 2x_2 &- x_3 + x_4 = 6, \\E_4 : \quad &\quad 2x_3 + 4x_4 = 12.\end{aligned}$$

- ▶ Exchange E_2 and E_3 : $(E_2) \leftrightarrow (E_3)$ (**pivoting**)

$$\begin{aligned}E_1 : \quad x_1 &- x_2 + 2x_3 - x_4 = -8, \\E_2 : \quad 2x_2 &- x_3 + x_4 = 6, \\E_3 : \quad - x_3 &- x_4 = -4, \\E_4 : \quad &\quad 2x_3 + 4x_4 = 12.\end{aligned}$$

Solving Linear Equations with pivoting (III)



$$\begin{aligned}E_1 : \quad x_1 &- x_2 + 2x_3 - x_4 = -8, \\E_2 : \quad &2x_2 - x_3 + x_4 = 6, \\E_3 : \quad &- x_3 - x_4 = -4, \\E_4 : \quad &2x_3 + 4x_4 = 12.\end{aligned}$$

- ▶ eliminate x_3 from E_4 : $(E_4 + E_3) \rightarrow (E_4)$

$$\begin{aligned}E_1 : \quad x_1 &- x_2 + 2x_3 - x_4 = -8, \\E_2 : \quad 2x_2 &- x_3 + x_4 = 6, \\E_3 : \quad &- x_3 - x_4 = -4, \\E_4 : \quad &\qquad\qquad\qquad 2x_4 = 4.\end{aligned}$$

- ▶ backward substitution:

$$x_4 = 2, \quad x_3 = 2, \quad x_2 = 3, \quad x_1 = -7.$$

System of 2nd order ODEs

System of m second-order ODEs:

$$\frac{d^2 u_1}{dt^2} = f_1(t, u_1, u'_1, u_2, u'_2, \dots, u_m, u'_m),$$

$$\frac{d^2 u_2}{dt^2} = f_2(t, u_1, u'_1, u_2, u'_2, \dots, u_m, u'_m),$$

⋮

$$\frac{d^2 u_m}{dt^2} = f_m(t, u_1, u'_1, u_2, u'_2, \dots, u_m, u'_m), \quad a \leq t \leq b,$$

with $2m$ initial conditions

$$u_1(a) = \alpha_1, \quad u'_1(a) = \alpha'_1, \quad u_2(a) = \alpha_2, \quad u'_2(a) = \alpha'_2, \dots,$$

$$u_m(a) = \alpha_m, \quad u'_m(a) = \alpha'_m.$$

u_1, u_2, \dots, u_m could be vectors.

System of 2nd order ODEs: first-order form

$$\mathbf{z} = \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_{2m} \end{pmatrix} \stackrel{\text{def}}{=} \begin{pmatrix} u_1 \\ u'_1 \\ u_2 \\ u'_2 \\ \vdots \\ u_m \\ u'_m \end{pmatrix}, \quad \mathbf{f}(t, \mathbf{z}) \stackrel{\text{def}}{=} \begin{pmatrix} z_2 \\ f_1(t, z_1, z_2, z_3, \dots, z_{2m-1}, z_{2m}) \\ z_4 \\ f_2(t, z_1, z_2, z_3, \dots, z_{2m-1}, z_{2m}) \\ \vdots \\ z_{2m} \\ f_m(t, z_1, z_2, z_3, \dots, z_{2m-1}, z_{2m}) \end{pmatrix}$$

System of $2m$ first-order ODEs: $\frac{d\mathbf{z}}{dt} = \mathbf{f}(t, \mathbf{z}), \quad a \leq t \leq b,$

with initial condition $\mathbf{z}(a) = \boldsymbol{\alpha} \stackrel{\text{def}}{=} \begin{pmatrix} \alpha_1 \\ \alpha'_1 \\ \alpha_2 \\ \alpha'_2 \\ \vdots \\ \alpha_m \\ \alpha'_m \end{pmatrix}.$