

# ESS101- Modeling and Simulation

## Lecture 18-19

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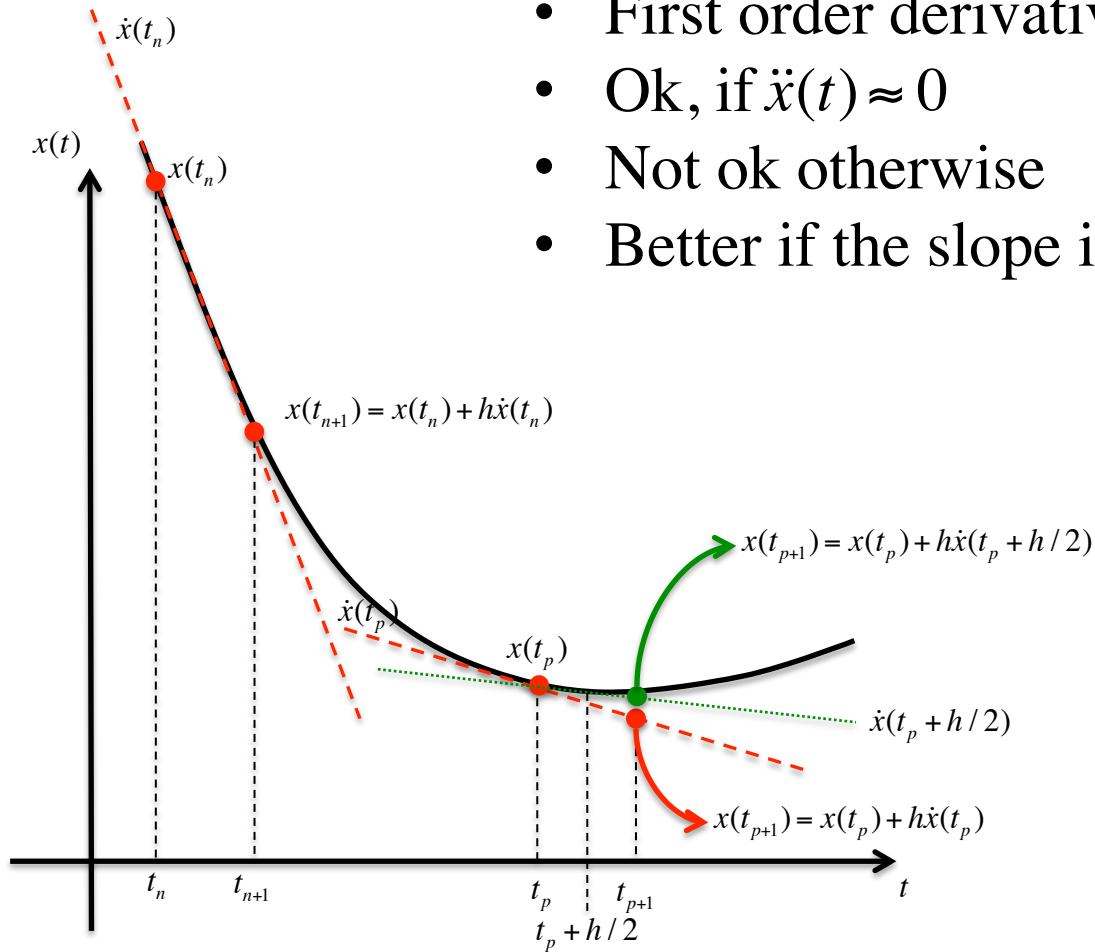
# Today (Chapters 9, 10)

- ➡ Motivating example
- ➡ General RK methods
- ➡ Four multi-stage RK methods
- ➡ Absolute stability of RK methods

# Motivating example

## Remarks

- First order derivative used
- Ok, if  $\ddot{x}(t) \approx 0$
- Not ok otherwise
- Better if the slope is “sampled” finer



# Numerical Example

Simulate the system

$$\dot{x}(t) = (1 - 2t)x(t), \quad x(0) = 1$$

$$0 \leq t \leq 1.2$$

Recall the *exact solution*  $x(t) = e^{\frac{1}{4} - \left(\frac{1}{2} - t\right)^2}$

Calculate the approximate solution as

$$x_{n+1} = x_n + h\dot{x}_n \quad (\text{FE}), \rightarrow x_{n+1} = x_n + hk_2,$$

$$k_2 = \dot{x}\left(t_n + \frac{1}{2}h\right) = (1 - 2t_n - h)x\left(t_n + \frac{1}{2}h\right)$$

$$x\left(t_n + \frac{1}{2}h\right) \approx x_n + \frac{h}{2}\dot{x}_n$$

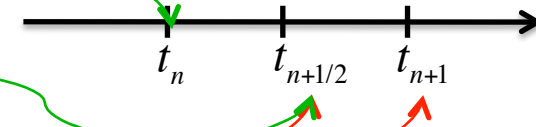
## Numerical Example

The  $n$ -th iteration is

$$k_1 = f(t_n, x_n) \text{ (slope)}$$

$$k_2 = f\left(t_n + \frac{1}{2}h, x_n + \frac{1}{2}hk_1\right) \text{ (slope)}$$

$$x_{n+1} = x_n + hk_2, \text{ (solution)}$$



GE\*10<sup>3</sup>@t=1.2


$h$	TS(2)	Trap.	ABE	ABT	RK(2)
0.2	5.4	-2.8	-3.6	17.6	3.5
0.1	1.4	-0.71	-0.66	4.0	0.67
Ratio	3.90	4.00	5.49	4.40	5.24

2<sup>nd</sup> order methods


# General RK methods

The  $n$ -th iteration of a ***s-stage*** method is

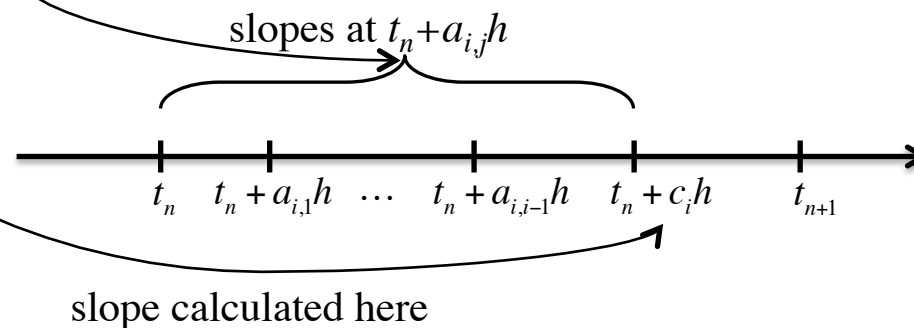
$$x_{n+1} = x_n + h \sum_{i=1}^s b_i k_i, \quad \text{Weighted linear combination of slopes in } [t_n, t_{n+1}]$$



$$k_i = f\left(t_n + c_i h, x_n + h \sum_{j=1}^s a_{i,j} k_j\right), \quad i = 1, \dots, s$$



Pick points in  $[t_n, t_{n+1}]$  where slopes are calculated



Hence, natural choices for  $c_i$  and  $a_{i,j}$  are

$$c_i = \sum_{j=1}^s a_{i,j}, \quad i = 1, \dots, s \qquad a_{i,j} = 0, \quad j \geq i$$

# General RK methods

$c_1$	$a_{1,1}$	$a_{1,2}$	$\cdots$	$a_{1,s}$
$c_2$	$a_{2,1}$	$a_{2,2}$	$\cdots$	$a_{2,s}$
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$c_s$	$a_{s,1}$	$a_{s,2}$	$\cdots$	$a_{s,s}$
	$b_1$	$b_2$	$\cdots$	$b_s$

*Butcher array* for a  
full RK method  
(implicit)

0	0	0	$\cdots$	0
$a_{2,1} + a_{2,2}$	$a_{2,1}$	$a_{2,2}$	$\cdots$	0
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$\sum_{j=1}^s a_{s,j}$	$a_{s,1}$	$a_{s,2}$	$\cdots$	$a_{s,s}$
	$b_1$	$b_2$	$\cdots$	$b_s$

*Butcher array* for an  
explicit RK method

For a  $s$ -stage method  $s^2+s$  parameters  $(a_{i,j}, b_i)$  must be determined. How?

With fixed  $s$ , maximize the method order

# One-stage methods

A  $s$ -stage method is of order  $p$  if

$$e_{n+1} = x(t_{n+1}) - x_{n+1} = O(h^{p+1})$$

Recall the  $e_{n+1}$  is the local error. That is, calculated by assuming  $x(t_n) = x_n$

$$\begin{aligned} \text{For } s=1 \quad x_{n+1} &= x_n + hb_1k_1 & k_1 &= f(t_n + c_1h, x_n + a_{1,1}k_1) \\ & & &= \left. f(t_n, x_n) \right|_{\substack{c_1=0 \\ a_{1,1}=0}} \end{aligned}$$

To form the local error let's expand  $x(t_{n+1})$  is Taylor series

$$x(t_{n+1}) = x(t_n) + h\dot{x}(t_n) + \frac{1}{2!}h^2\ddot{x}(t_n) + O(h^3)$$

where  $\ddot{x}(t_n)$  is calculated as  $\ddot{x}(t_n) = \frac{\partial}{\partial t} f(t_n, x_n) + \dot{x}(t_n) \frac{\partial}{\partial x} f(t_n, x_n)$



## One-stage methods

$$\text{Hence } x(t_{n+1}) = x(t_n) + h\dot{x}(t_n) + \frac{1}{2!}h^2 \left[ f_t + \dot{x}(t_n)f_x \right] + O(h^3)$$

The expression of the local error is

$$\begin{aligned} e_{n+1} &= x(t_{n+1}) - x_{n+1} \\ &= x(t_n) + h\dot{x}(t_n) + \frac{1}{2!}h^2 \left[ f_t + \dot{x}(t_n)f_x \right] - x(t_n) - hb_1 \overbrace{f(t_n, x(t_n))}^{\dot{x}(t_n)} + O(h^3) \\ &= h(1 - b_1)\dot{x}(t_n) + \frac{1}{2!}h^2 \left[ f_t + \dot{x}(t_n)f_x \right] + O(h^3) \end{aligned}$$

The maximum attainable order is  $p=1$  with  $b_1=1$

$$x_{n+1} = x_n + hk_1 \quad k_1 = f(t_n, x_n) \quad \textbf{\textit{FE Euler method}}$$

## Two-stage methods

For  $s=2$  the  $n$ -th iteration is

$$\begin{aligned}x_{n+1} &= x_n + h(b_1 k_1 + b_2 k_2), \\k_1 &= f(t_n, x_n) \\k_2 &= f(t_n + ah, x_n + ahk_1)\end{aligned}$$

Find  $a, b_1, b_2$  to maximize the method order

To calculate the **local error** write the method with  $x_n = x(t_n)$

$$\begin{aligned}x_{n+1} &= x(t_n) + h(b_1 f(t_n, x(t_n)) + b_2 f(t_n + ah, x(t_n) + ahf(t_n, x(t_n)))) \\&= x(t_n) + hb_1 \dot{x}(t_n) + hb_2 f(t_n + ah, x(t_n) + ah\dot{x}(t_n)) \\&= x(t_n) + hb_1 \dot{x}(t_n) + hb_2 \left( f(t_n, x(t_n)) + ah \frac{\partial}{\partial t} f(t_n, x(t_n)) + ah\dot{x}(t_n) \frac{\partial}{\partial x} f(t_n, x(t_n)) + O(h^2) \right) \\&= x(t_n) + hb_1 \dot{x}(t_n) + hb_2 (\dot{x}(t_n) + ahf_t + ah\dot{x}(t_n)f_x + O(h^2))\end{aligned}$$

## Two-stage methods

To calculate the local error expand  $x(t_{n+1})$  as

$$x(t_{n+1}) = x(t_n) + h\dot{x}(t_n) + \frac{1}{2!}h^2[f_t + \dot{x}(t_n)f_x] + \frac{1}{3!}h^3\ddot{x}(t_n) + O(h^4)$$

Calculate the local error  $e_{n+1}$  as

$$\begin{aligned} e_{n+1} &= x(t_{n+1}) - x_{n+1} \\ &= x(t_n) + h\dot{x}(t_n) + \frac{1}{2!}h^2[f_t + \dot{x}(t_n)f_x] + \frac{1}{3!}h^3\ddot{x}(t_n) + O(h^4) \\ &\quad - x(t_n) - h(b_1 + b_2)\dot{x}(t_n) - hb_2(ahf_t + ah\dot{x}(t_n)f_x + O(h^2)) \\ &= h(1 - b_1 - b_2)\dot{x}(t_n) + h^2\left(\frac{1}{2} - ab_2\right)[f_t + \dot{x}(t_n)f_x] + \frac{1}{3!}h^3\ddot{x}(t_n) + O(h^4) \end{aligned}$$

## Two-stage methods

$$e_{n+1} = h(1 - b_1 - b_2)\dot{x}(t_n) + h^2\left(\frac{1}{2} - ab_2\right)\left[f_t + \dot{x}(t_n)f_x\right] + \frac{1}{3!}h^3\ddot{x}(t_n) + O(h^4)$$

By setting  $b_1 + b_2 = 1$  the method has order  **$p=1$**   $\forall a$

By setting  $b_1 + b_2 = 1$  and  $ab_2 = \frac{1}{2}$  the method has order  **$p=2$**

*Two conditions on three parameters for **order two***

**Q:** Can one more condition be derived by expanding the method equations further to achieve order 3?

**A:** It can be shown (see p. 129) that there is not a combination of parameter enabling  $p=3$

# Three-stage methods

For  $s=3$  the  $n$ -th iteration is

$$x_{n+1} = x_n + h(b_1 k_1 + b_2 k_2 + b_3 k_3), \quad \begin{aligned} k_2 &= f(t_n + c_2 h, x_n + a_{2,1} h k_1) \\ k_3 &= f(t_n + c_3 h, x_n + a_{3,1} h k_1 + a_{3,2} h k_2) \\ k_1 &= f(t_n, x_n) \end{aligned}$$

Similarly to 2-stage methods, it can be shown that

$$b_1 + b_2 + b_3 = 1 \quad (\text{order 1})$$

$$b_2 c_2 + b_3 c_3 = \frac{1}{2} \quad (\text{order 2})$$

$$\left. \begin{aligned} b_2 c_2^2 + b_3 c_3^2 &= \frac{1}{3} \\ c_2 a_{3,2} b_3 &= \frac{1}{6} \end{aligned} \right\} \quad (\text{order 3})$$

$$\begin{array}{c|ccc} 0 & 0 & & \\ \frac{1}{3} & \frac{1}{3} & 0 & \\ \frac{2}{3} & 0 & \frac{2}{3} & 0 \\ \hline & \frac{1}{4} & 0 & \frac{3}{4} \end{array}$$

Heun's 3<sup>rd</sup> order rule

$$\begin{array}{c|ccc} 0 & 0 & & \\ \frac{1}{2} & \frac{1}{2} & 0 & \\ \frac{1}{2} & -1 & 2 & 0 \\ \hline & \frac{1}{6} & \frac{4}{3} & \frac{1}{6} \end{array}$$

RK's 3<sup>rd</sup> order rule

# Four-stage methods

0	0			
$\frac{1}{2}$	$\frac{1}{2}$	0		
$\frac{1}{2}$	0	$\frac{1}{2}$	0	
1	0	0	1	0
	$\frac{1}{6}$	$\frac{1}{3}$	$\frac{1}{3}$	$\frac{1}{6}$

Order  $s$  cannot be achieved for a  $s$ -step method with  $s > 4$

Higher step methods are impractical. Large number of equations to solve.

# Absolute stability

**Definition (Absolute stability).** A RK method is **absolutely stable** if its solution  $x_n$  to the problem

$$\dot{x}(t) = \lambda x(t), \quad \operatorname{Re}(\lambda) < 0$$

converges to zero as  $n \rightarrow \infty$

## Example

Study the stability of a RK(2) with the following coefficients

$$\begin{aligned} x_{n+1} &= x_n + h(b_1 k_1 + b_2 k_2), \\ k_1 &= f(t_n, x_n) \\ k_2 &= f(t_n + ah, x_n + ahk_1) \end{aligned}$$

$$\begin{array}{c|cc} 0 & 0 & \\ \frac{1}{2\theta} & \frac{1}{2\theta} & \\ \hline & 1-\theta & \theta \end{array} \quad (\text{order 2 conditions})$$

# Absolute stability

$$\begin{aligned}x_{n+1} &= x_n + h(b_1 k_1 + b_2 k_2), \\k_1 &= f(t_n, x_n) \\k_2 &= f(t_n + ah, x_n + ahk_1)\end{aligned}$$

$$\begin{array}{c|cc} 0 & 0 & \\ \frac{1}{2\theta} & \frac{1}{2\theta} & \\ \hline & 1-\theta & \theta \end{array}$$

The  $n$ -th iteration becomes

$$\begin{aligned}x_{n+1} &= x_n + h(1-\theta)\lambda x_n + h\theta\left(\lambda x_n + \frac{\lambda^2 h}{2\theta} x_n\right) \\&= \left(1 + h\lambda + \frac{\lambda^2 h^2}{2}\right)x_n\end{aligned}$$

The corresponding characteristic polynomial is

$$p(r) = r - \left(1 + h\lambda + \frac{\lambda^2 h^2}{2}\right) \Rightarrow \left|1 + h\lambda + \frac{\lambda^2 h^2}{2}\right| < 1 \quad \text{Stability condition}$$



# Stability regions

