

수치해석 기초

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Contents

- PDE basics
- Numerical approximation of PDE
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- Methods
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 - Spectral method (*Jul 28th)
- Convergence

Partial differential equations (PDEs)

PDEs are common in physics, describing change of a system over space and time

e.g. Discharging RC circuit

$$\frac{dQ}{dt} = -\frac{1}{RC}Q$$

- unknown solution : Q \longrightarrow *dependent variable*
- is a function of : t \longrightarrow *independent variable*

e.g. 1D wave equation

$$\frac{\partial^2 u}{\partial t^2} = c^2 \frac{\partial^2 u}{\partial x^2}$$

- unknown solution : u
- is a function of : (x, t)

Differential equations with...

- 1 independent var : **ordinary differential equation (ODE)**
- ≥ 2 independent vars : **partial differential equation (PDE)**

Partial differential equations (PDEs)

e.g. Schrodinger equation in 1D

$$i\hbar \frac{\partial \Psi}{\partial t} = -\frac{\hbar^2}{2m} \frac{\partial^2 \Psi}{\partial x^2} + V\Psi$$

PDE, 2 independent vars, 1 dependent var $\Psi(x, t)$

e.g. Lotka-Volterra equations (predator-prey model)

$$\frac{dx}{dt} = Ax - Bxy$$

$$\frac{dy}{dt} = Cxy - Dy$$

system of 2 ODEs (not PDE)

e.g. Einstein equations in vacuum

$$G_{\mu\nu} = 0$$

system of 10 PDEs,

4 independent vars, 10 dependent vars $g_{\mu\nu}(x^a)$

Partial differential equations (PDEs)

PDEs may permit analytic (exact) solutions

e.g. Wave equation (3D) $\frac{\partial^2 u}{\partial t^2} = c^2 \nabla^2 u$ \longrightarrow Plane wave
Spherical wave
...

e.g. vacuum Einstein equations $G_{\mu\nu} = 0$ \longrightarrow Flat (Minkowski) space
Schwarzschild geometry
...

however, in general situations, they need to be solved numerically

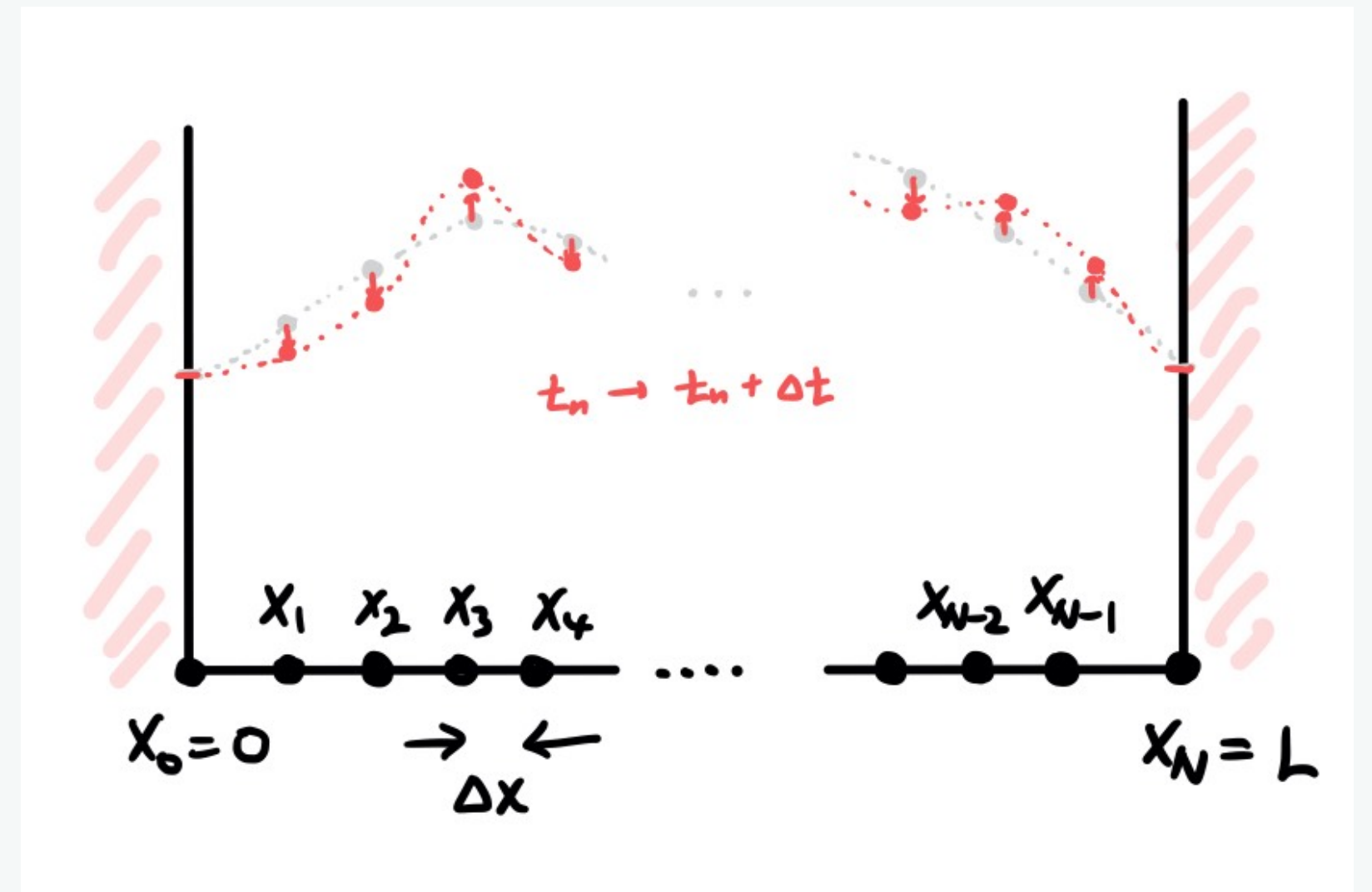
Example — heat equation

$$\frac{\partial u}{\partial t} = \alpha \frac{\partial^2 u}{\partial x^2}$$

- $x \in [0, L]$
- $u(x, 0) = u_0(x)$
- $u(0) = u(L) = 0$

Common practice :

- discretize computational domain
- approximate the PDE into discrete version
- apply initial condition
- evolve with time
- complete simulation



$$\frac{du_i}{dt} = \alpha \left(\frac{u_{i+1} - 2u_i + u_{i-1}}{(\Delta x)^2} \right) \equiv (*)$$

$$u_i = u_0(x_i)$$

$$u_i(t_{n+1}) = u_i(t_n) + (*) \times \Delta t$$

How to

- approximate spatial derivatives
- integrate with respect to time

Numerical derivative

How to approximate derivative on discrete grids?

$$f(x_0 + h) = f(x_0) + hf'(x_0) + \frac{h^2}{2}f''(x_0) + \dots$$

$$\frac{f(x_0 + h) - f(x_0)}{h} = f'(x_0) + \frac{h}{2}f''(x_0) + \dots$$

$$= f'(x_0) + \mathcal{O}(h)$$

truncate (ignore) higher order terms if h is small

truncation error

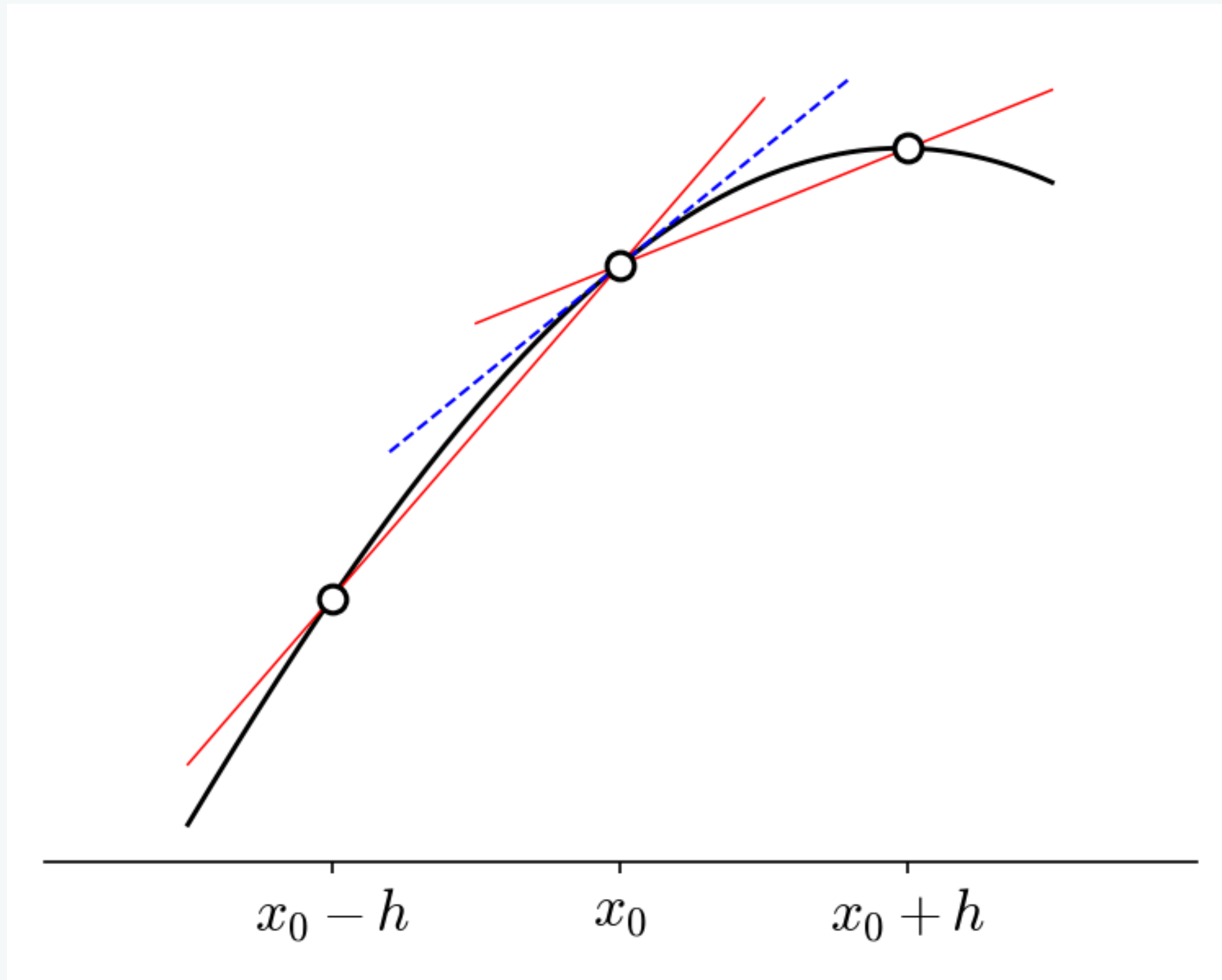
- Forward difference formula

$$\left. \frac{\partial f}{\partial x} \right|_{x_0} \approx \frac{f(x_0 + h) - f(x_0)}{h}$$

- Backward difference formula

$$\left. \frac{\partial f}{\partial x} \right|_{x_0} \approx \frac{f(x_0) - f(x_0 - h)}{h}$$

Numerical derivative



Error $\sim \mathcal{O}(h)$

first order accurate —
exact upto degree 1 polynomial (linear)

Numerical derivative

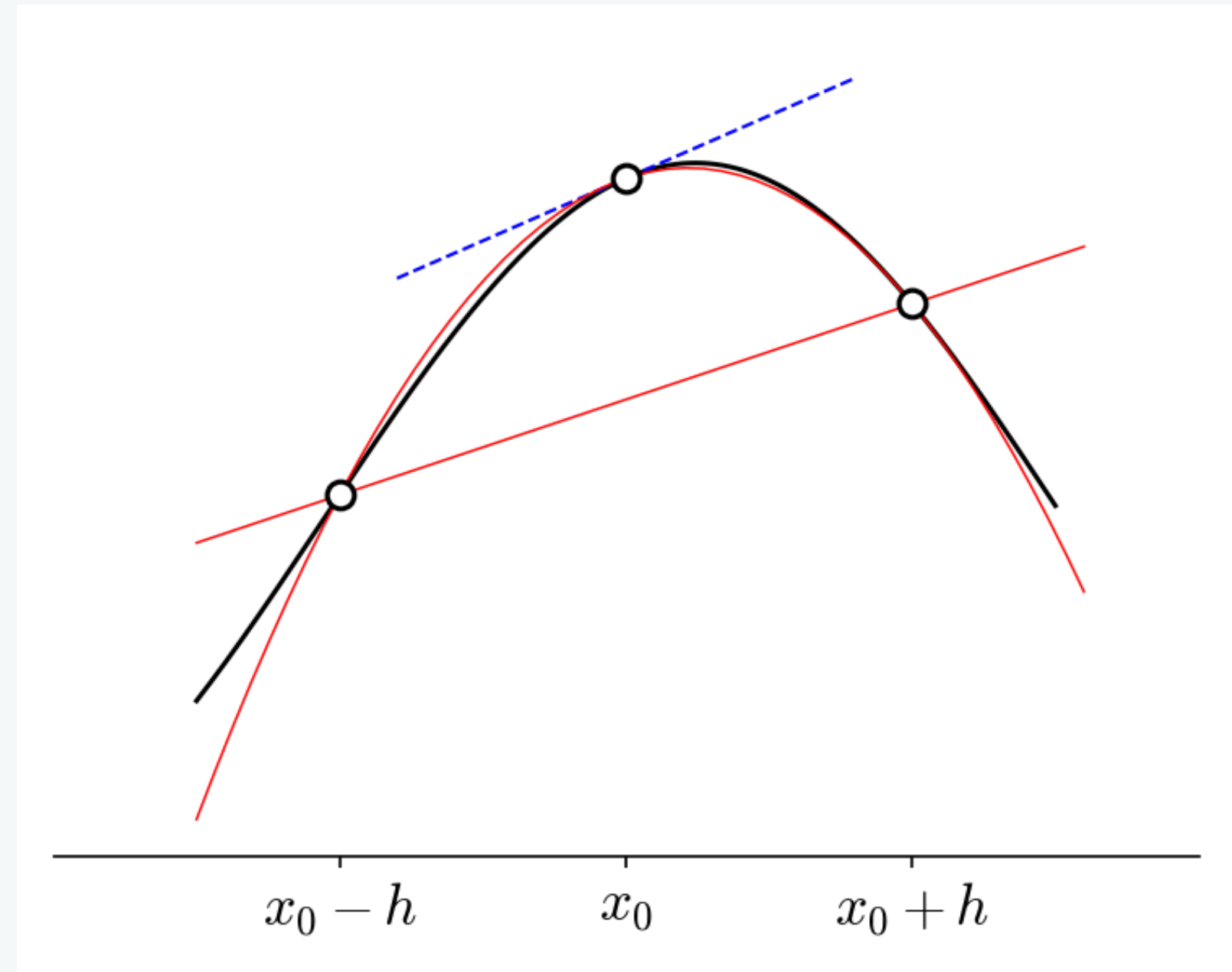
$$f(x_0 + h) = f(x_0) + hf'(x_0) + \frac{h^2}{2}f''(x_0) + \dots$$

$$f(x_0 - h) = f(x_0) - hf'(x_0) + \frac{h^2}{2}f''(x_0) + \dots$$

- Central difference (3-points) formula

$$\left. \frac{\partial f}{\partial x} \right|_{x_0} \approx \frac{f(x_0 + h) - f(x_0 - h)}{2h}$$

Error $\sim \mathcal{O}(h^2)$ (second order accurate)



Numerical derivative

Higher-order accuracy formula

- 5 points

$$f' = \frac{f(x_0 - 2h) - 8f(x_0 - h) + 8f(x_0 + h) - f(x_0 + 2h)}{12h} + \mathcal{O}(h^4)$$

- 7 points

$$f' = \frac{f(x_0 + 3h) - 9f(x_0 + 2h) + 45f(x_0 + h) - 45f(x_0 - h) + 9f(x_0 - 2h) - f(x_0 - 3h)}{60h} + \mathcal{O}(h^6)$$

note) more neighboring points ('stencil') are needed for higher order

Numerical derivative

Second derivatives

- 3-point

$$\left. \frac{\partial^2 f}{\partial x^2} \right|_{x_0} \approx \frac{f(x_0 + h) - 2f(x_0) + f(x_0 - h)}{h^2} + \mathcal{O}(h^2)$$

- 5-point

$$\left. \frac{\partial^2 f}{\partial x^2} \right|_{x_0} \approx \frac{-f(x_0 - 2h) + 16f(x_0 - h) - 30f(x_0) + 16f(x_0 + h) - f(x_0 + 2h)}{12h^2} + \mathcal{O}(h^4)$$

Numerical derivative

Round-off error

Is it always better to have smaller h ?

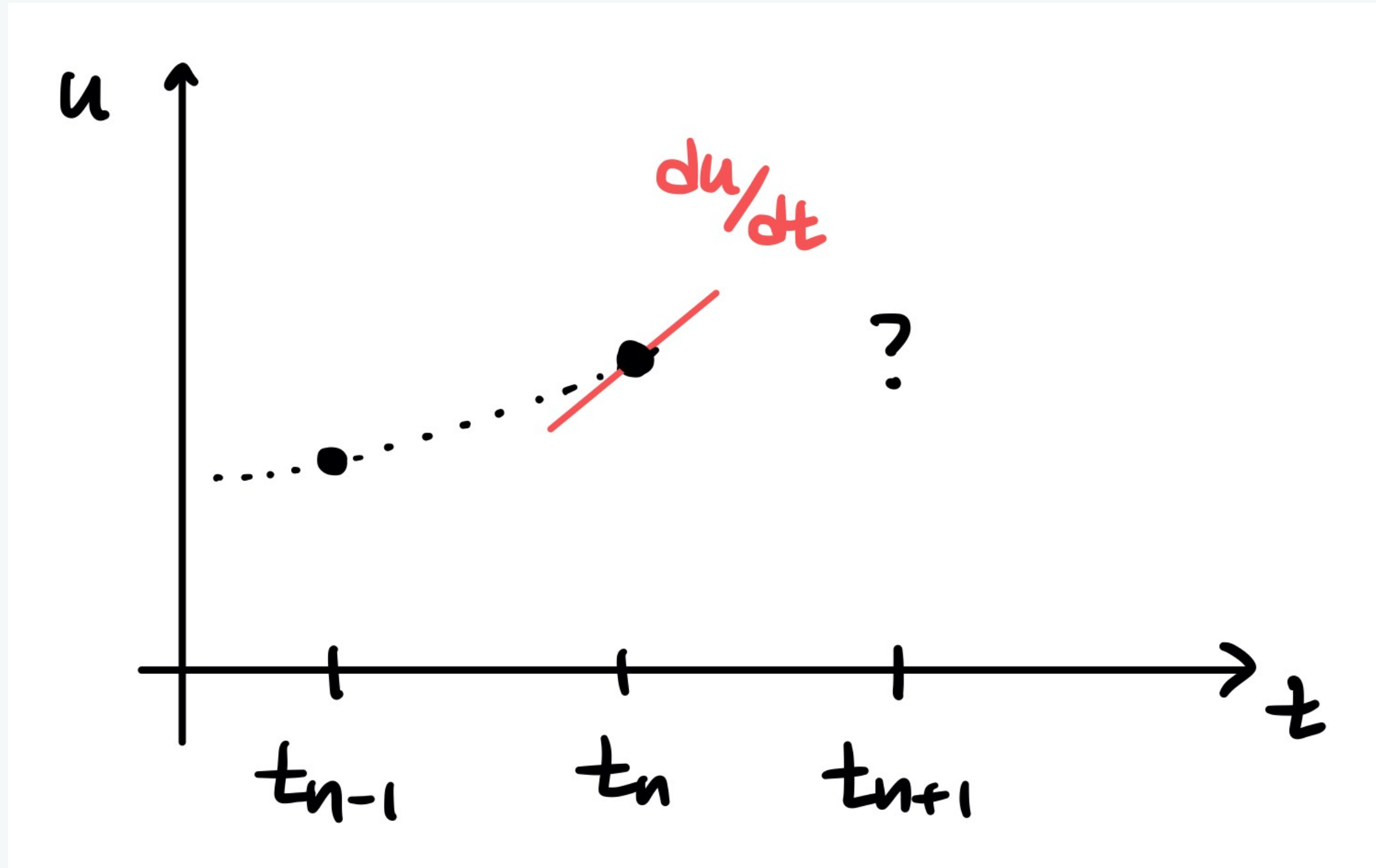
e.g. forward-difference numerical derivative of $f(x) = e^x$ at $x = 1$.

$$\frac{f(1+h) - f(1)}{h}$$

$$h = 10^{-k}, \quad k = 1, 2, \dots$$

for k = 1,	f'(1) = 2.858841954874,	error is 0.140560126415
for k = 2,	f'(1) = 2.731918655787,	error is 0.013636827328
for k = 3,	f'(1) = 2.719641422533,	error is 0.001359594074
for k = 4,	f'(1) = 2.718417747081,	error is 0.000135918622
for k = 5,	f'(1) = 2.718295419931,	error is 0.000013591472
for k = 6,	f'(1) = 2.718283187377,	error is 0.000001358918
for k = 7,	f'(1) = 2.718281965960,	error is 0.000000137501
for k = 8,	f'(1) = 2.718281825543,	error is -0.000000002916
for k = 9,	f'(1) = 2.718282054743,	error is 0.000000226284
for k = 10,	f'(1) = 2.718282053442,	error is 0.000000224983

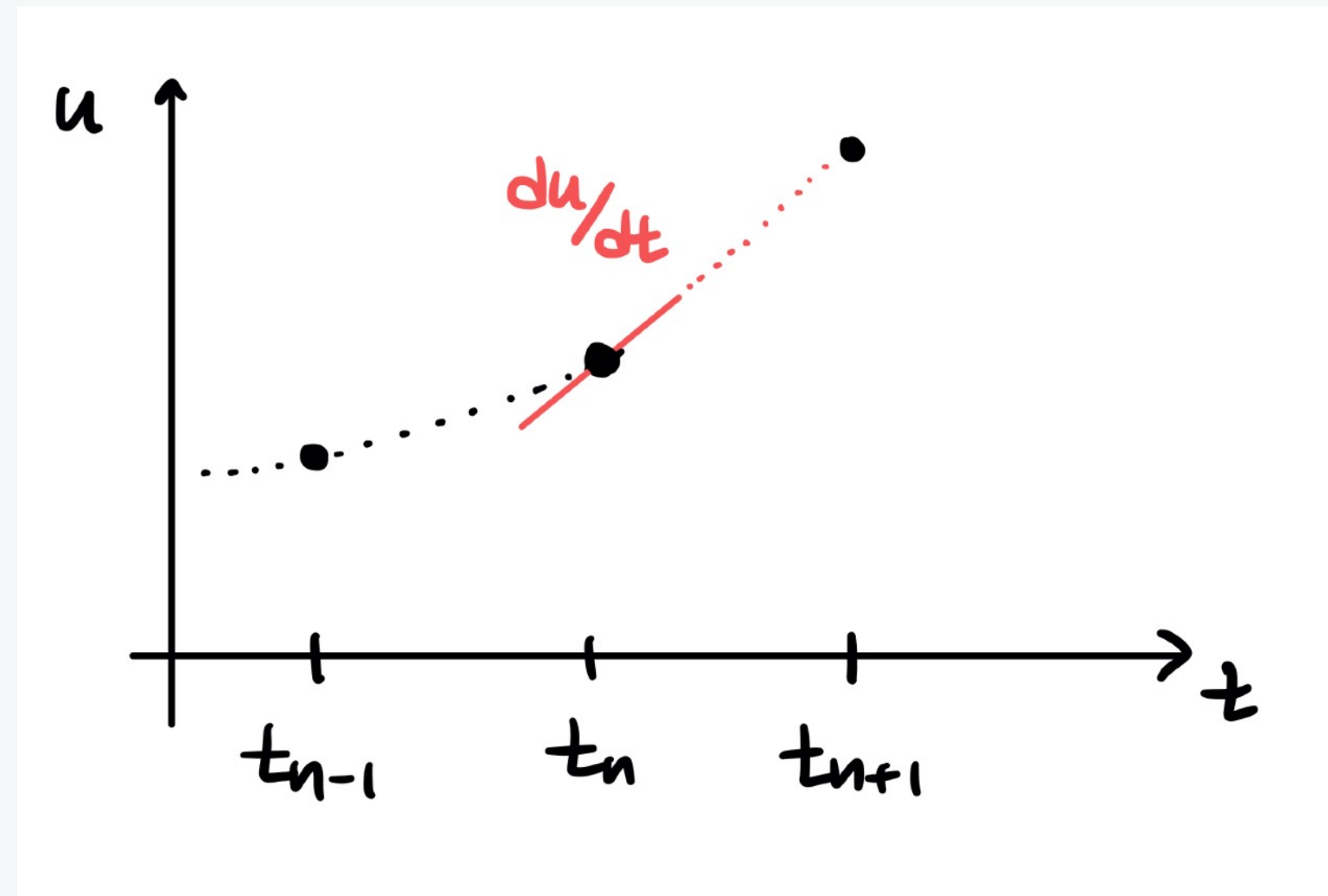
Time stepping



Time stepping

- Euler method (1st order)

$$u(x, t_{n+1}) \approx u(x, t_n) + \Delta t \left(\frac{\partial u}{\partial t} \right)_{(x, t_n)}$$

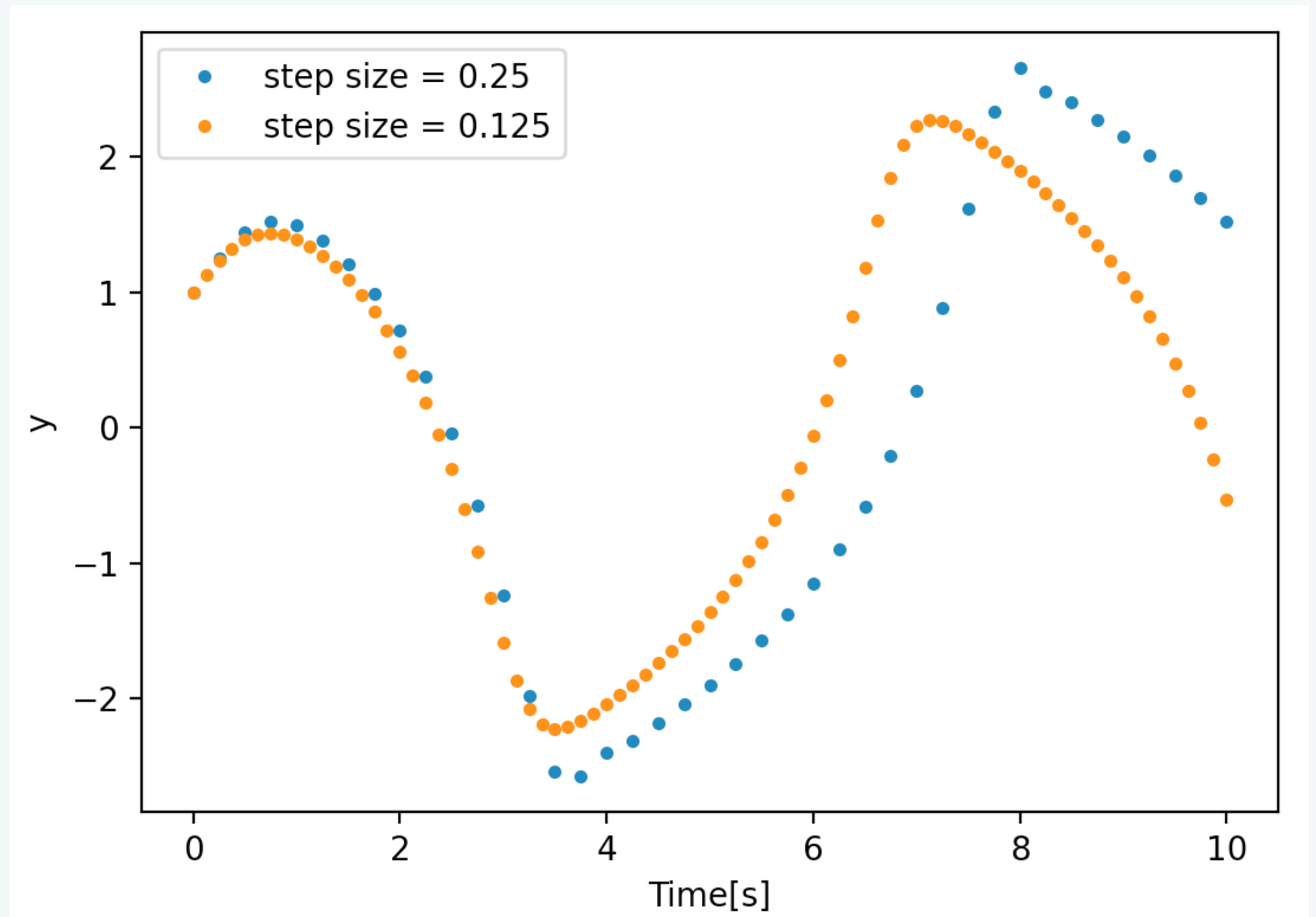


Time stepping

e.g. van der Pol equation

$$\frac{d^2y}{dt^2} - (1 - y^2)\frac{dy}{dt} + y = 0$$

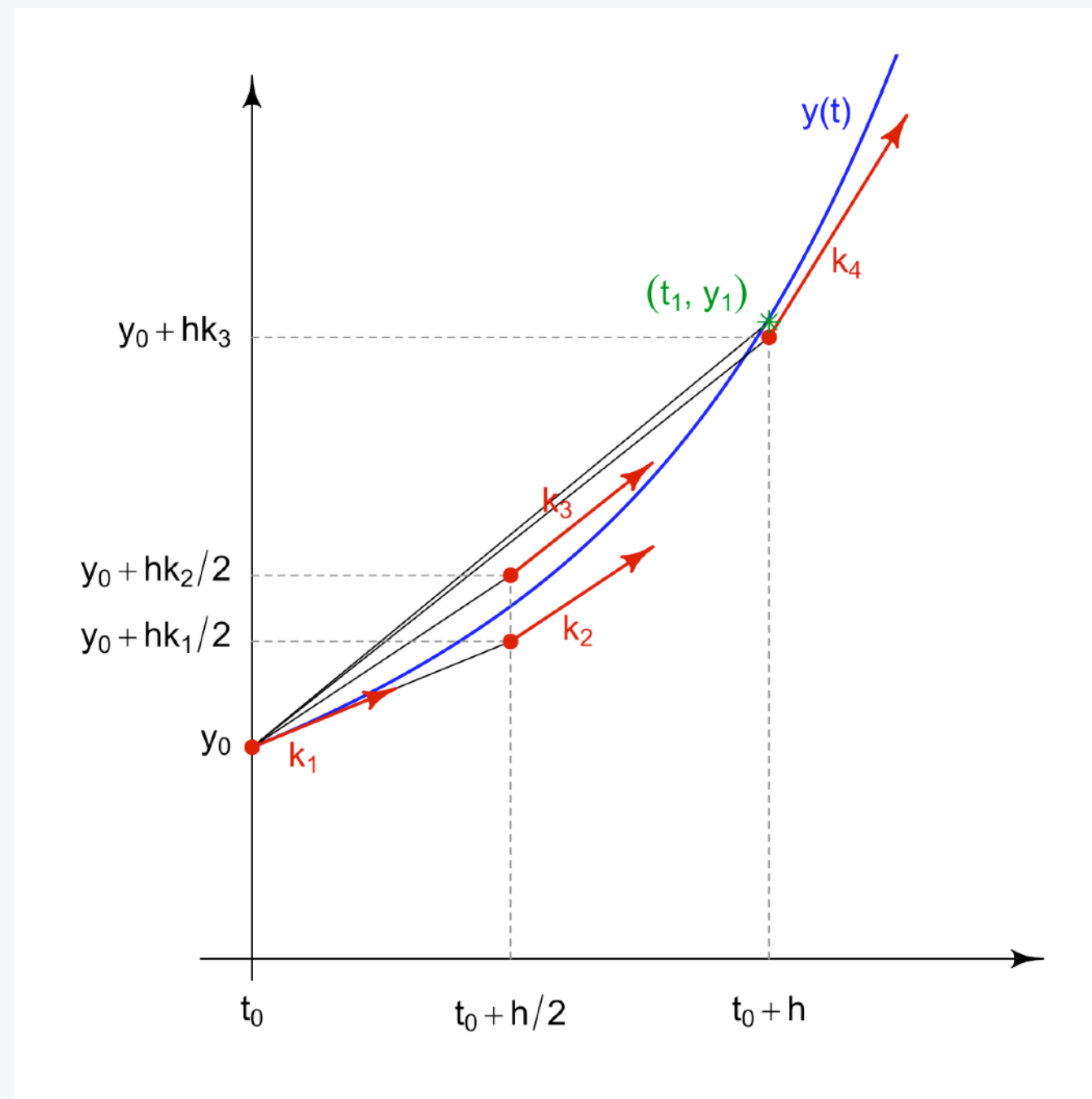
$$y(0) = y'(0) = 1$$



Time stepping

Higher order time stepping

e.g. Runge-Kutta methods



1 Explicit methods

- 1.1 Forward Euler
- 1.2 Explicit midpoint method
- 1.3 Heun's method
- 1.4 Ralston's method
- 1.5 Generic second-order method
- 1.6 Kutta's third-order method
- 1.7 Generic third-order method
- 1.8 Heun's third-order method
- 1.9 Van der Houwen's/Wray third-order method
- 1.10 Ralston's third-order method
- 1.11 Third-order Strong Stability Preserving Runge-Kutta (SSPRK3)
- 1.12 Classic fourth-order method
- 1.13 3/8-rule fourth-order method
- 1.14 Ralston's fourth-order method

2 Embedded methods

- 2.1 Heun-Euler
- 2.2 Fehlberg RK1(2)
- 2.3 Bogacki-Shampine
- 2.4 Fehlberg
- 2.5 Cash-Karp
- 2.6 Dormand-Prince

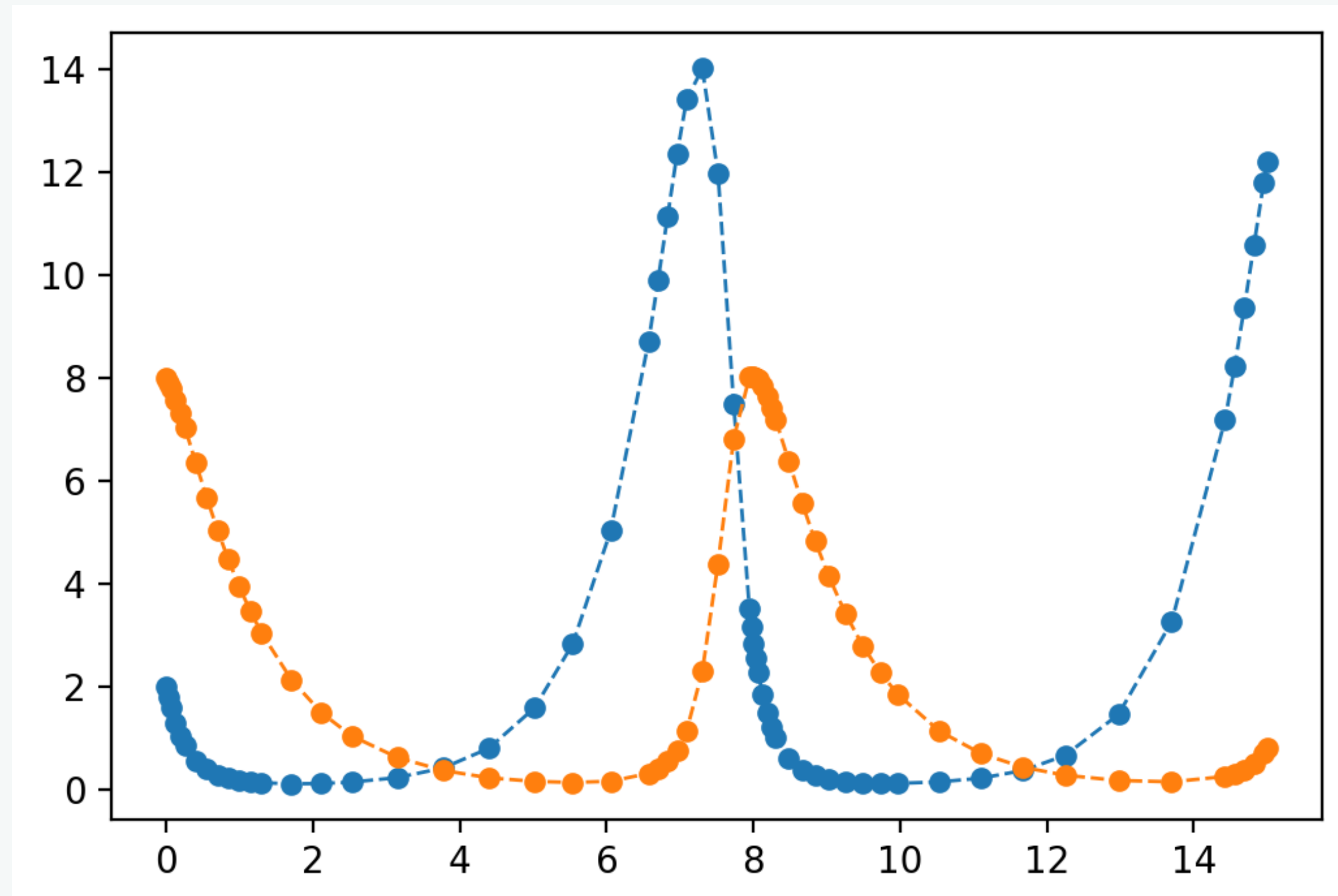
3 Implicit methods

- 3.1 Backward Euler
- 3.2 Implicit midpoint

Time stepping

Adaptive integrator

- changes step size with local truncation error

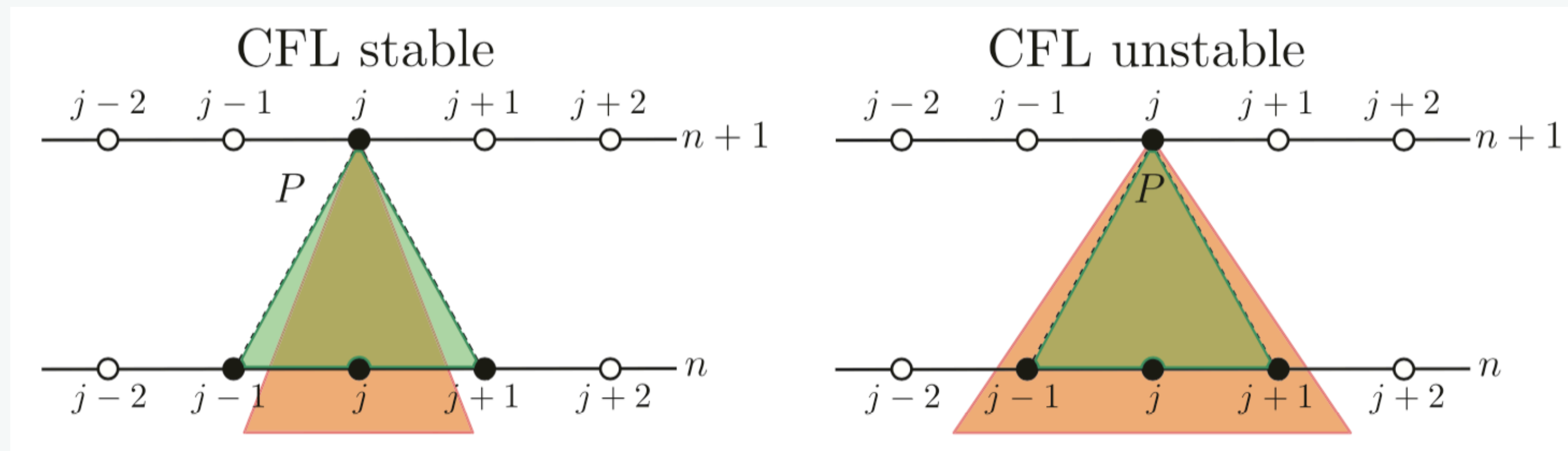


Time stepping

the Courant-Friedrichs-Lewy condition

- There is a max (limit) size of time step Δt
- CFL number : $\Delta t / \Delta t_{\max}$

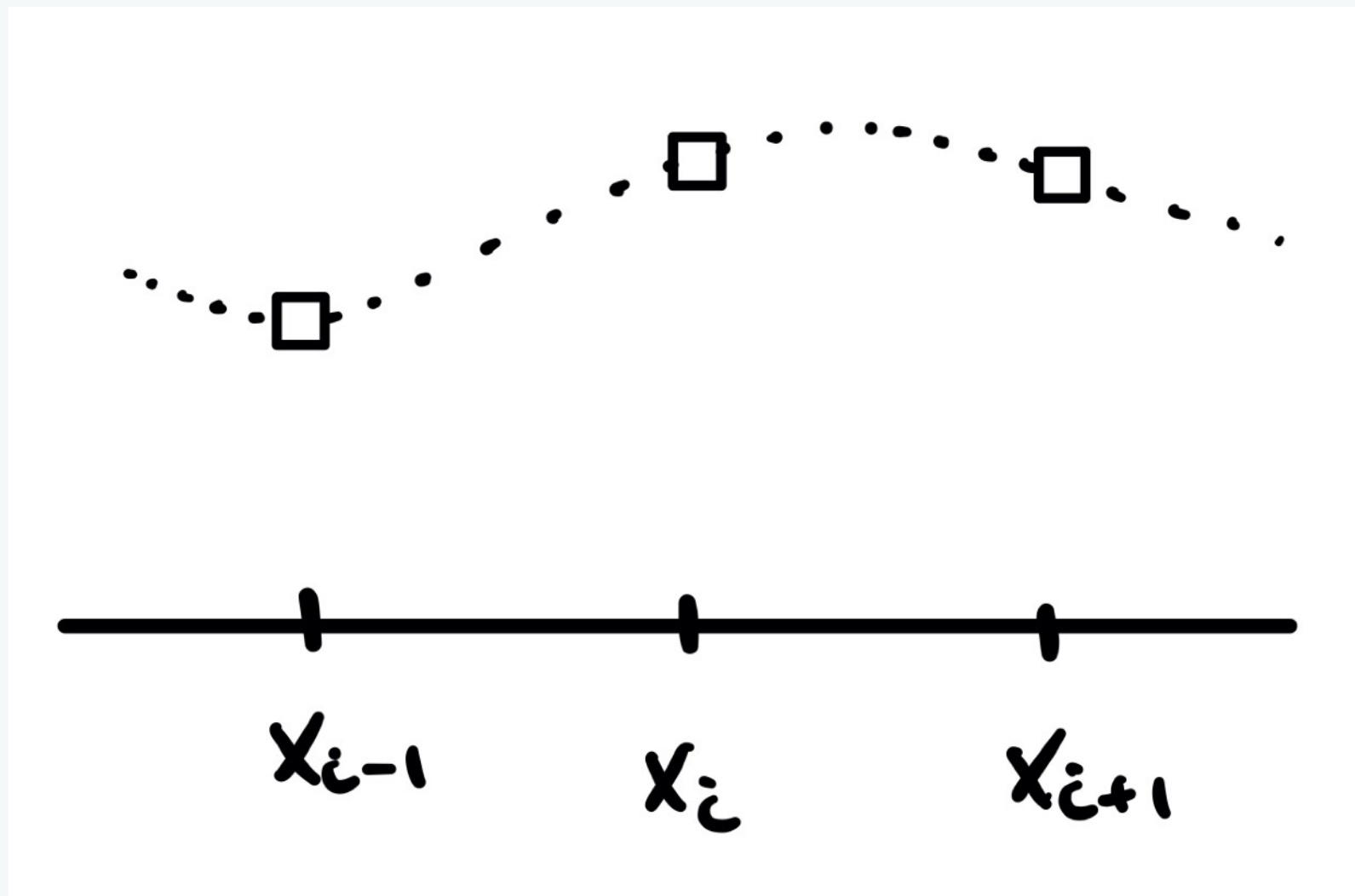
$$\Delta t_{\max} \sim \frac{C\Delta x}{\lambda} \quad \text{CFL limit}$$



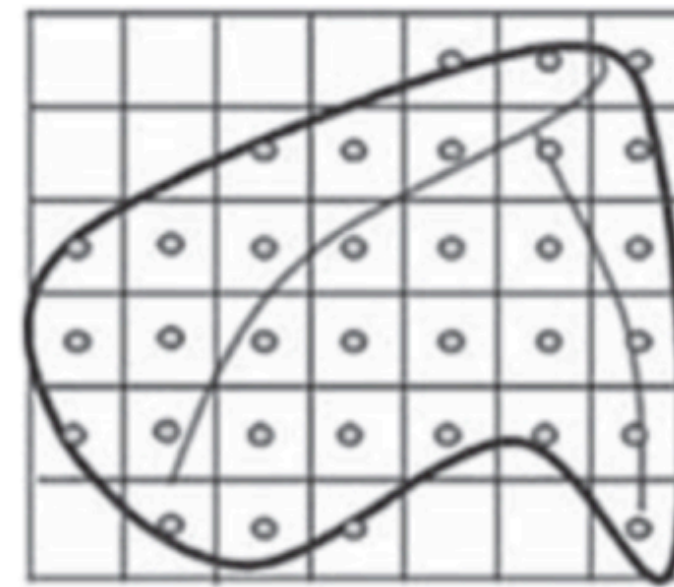
Rezzolla & Zanotti 2013

Finite difference method

Numerical values $\{u_i\}$ represent **pointwise values** at $x = x_i$



- Fast & easy to implement
- Hard to use for complex geometries e.g. non-cartesian



(a) Finite Differences

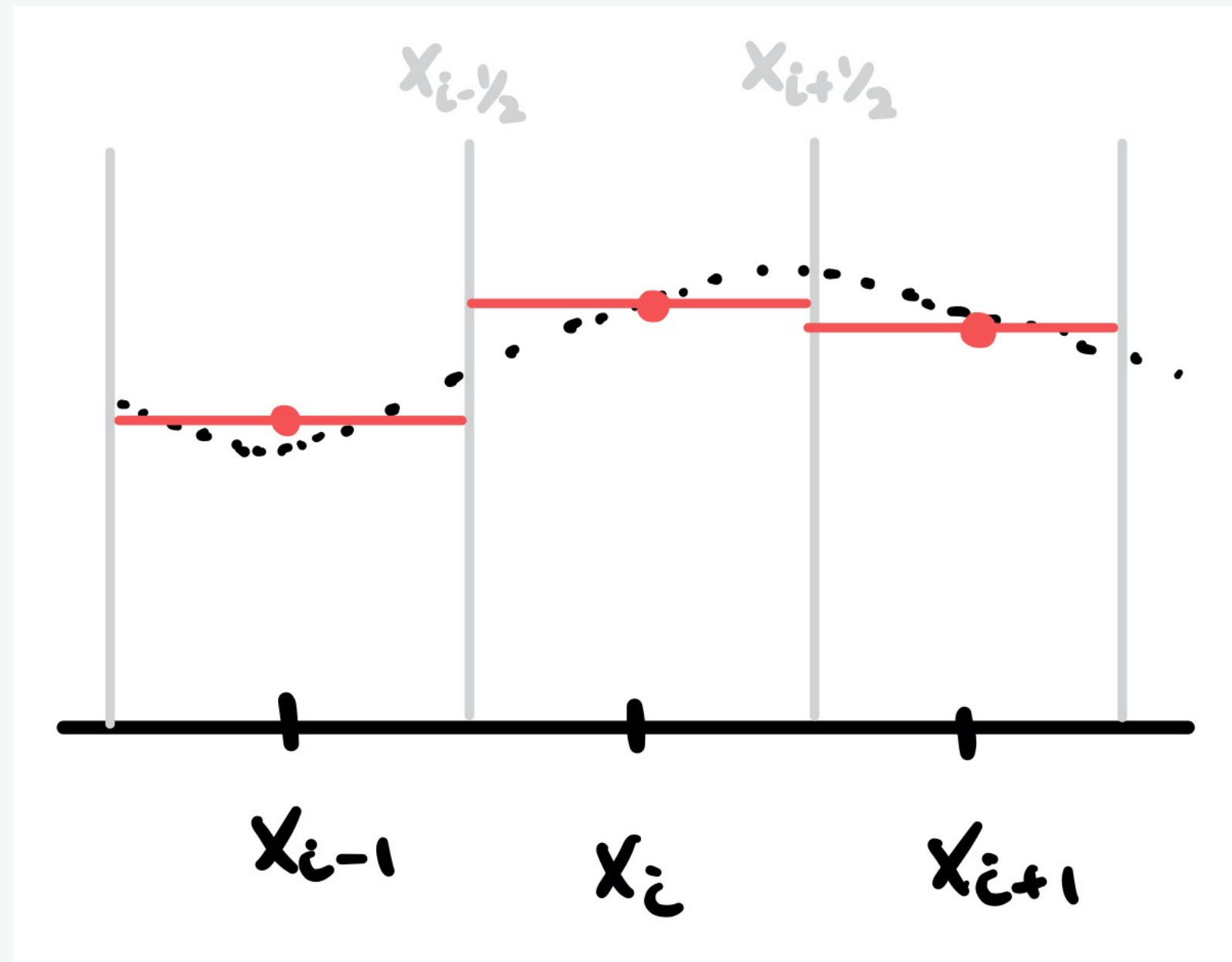


(b) Finite Volumes

Image credit : https://ebrary.net/215164/environment/finite_differences

Finite volume method

Numerical values $\{u_i\}$ represent **volume-averaged value** over the cell $[x_{i-1/2}, x_{i+1/2}]$

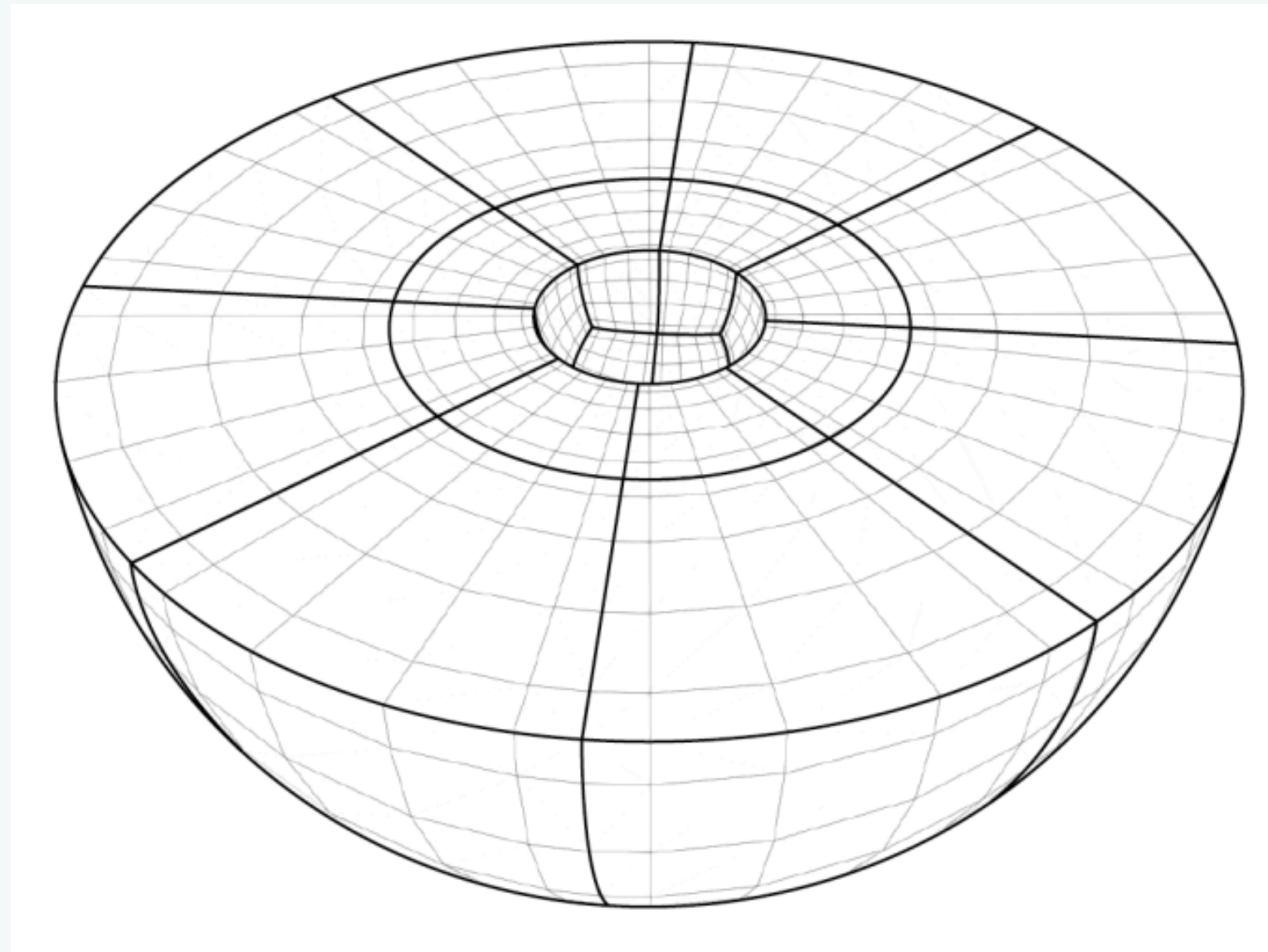


- Useful for conservation-type equations, Resolves discontinuities very well e.g. fluid dynamics
- can handle unstructured mesh
- computationally more expensive

Convergence of numerical methods

How many grid points do we need for desired accuracy?

- computational cost scales $\sim N^D$



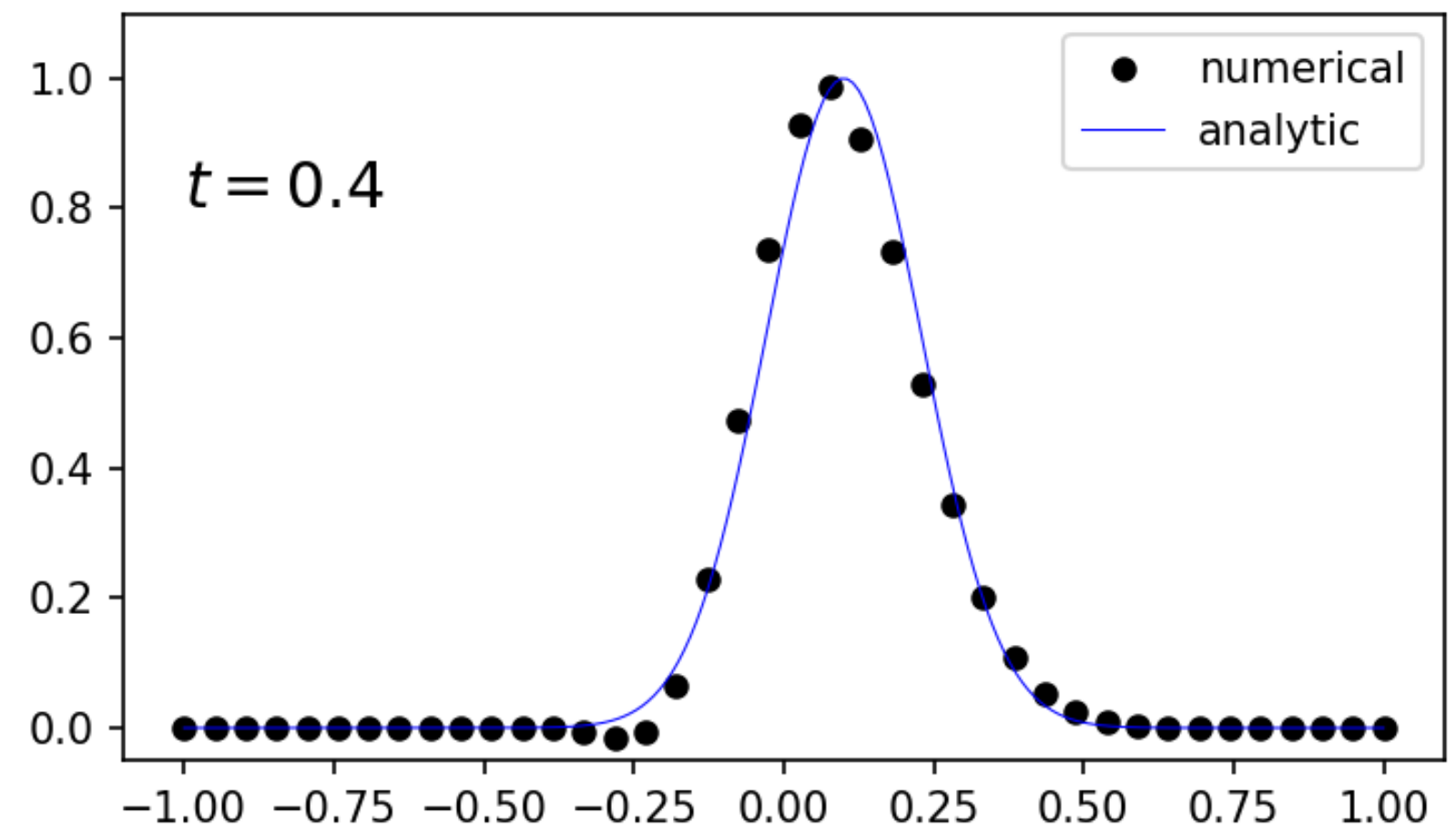
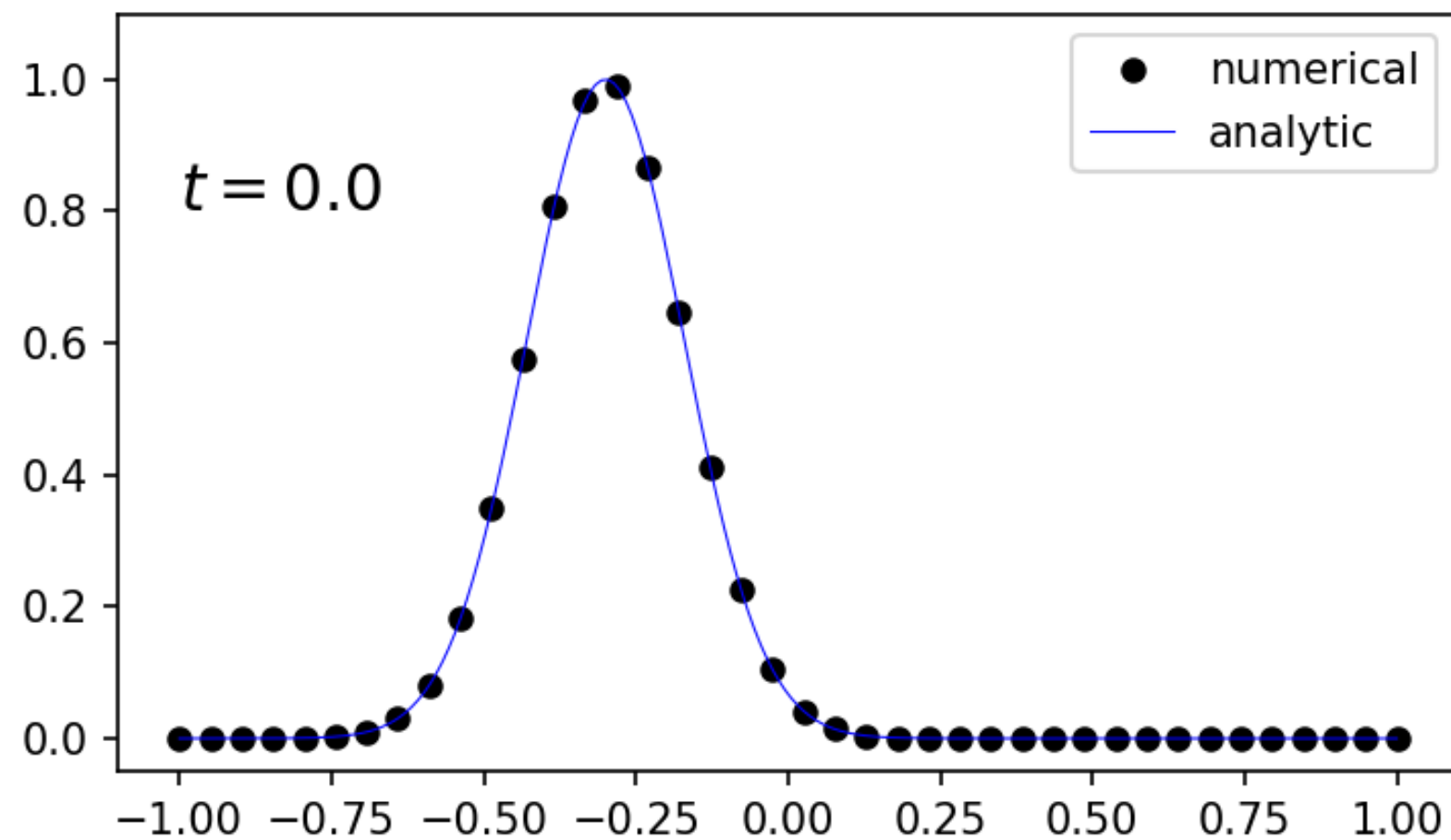
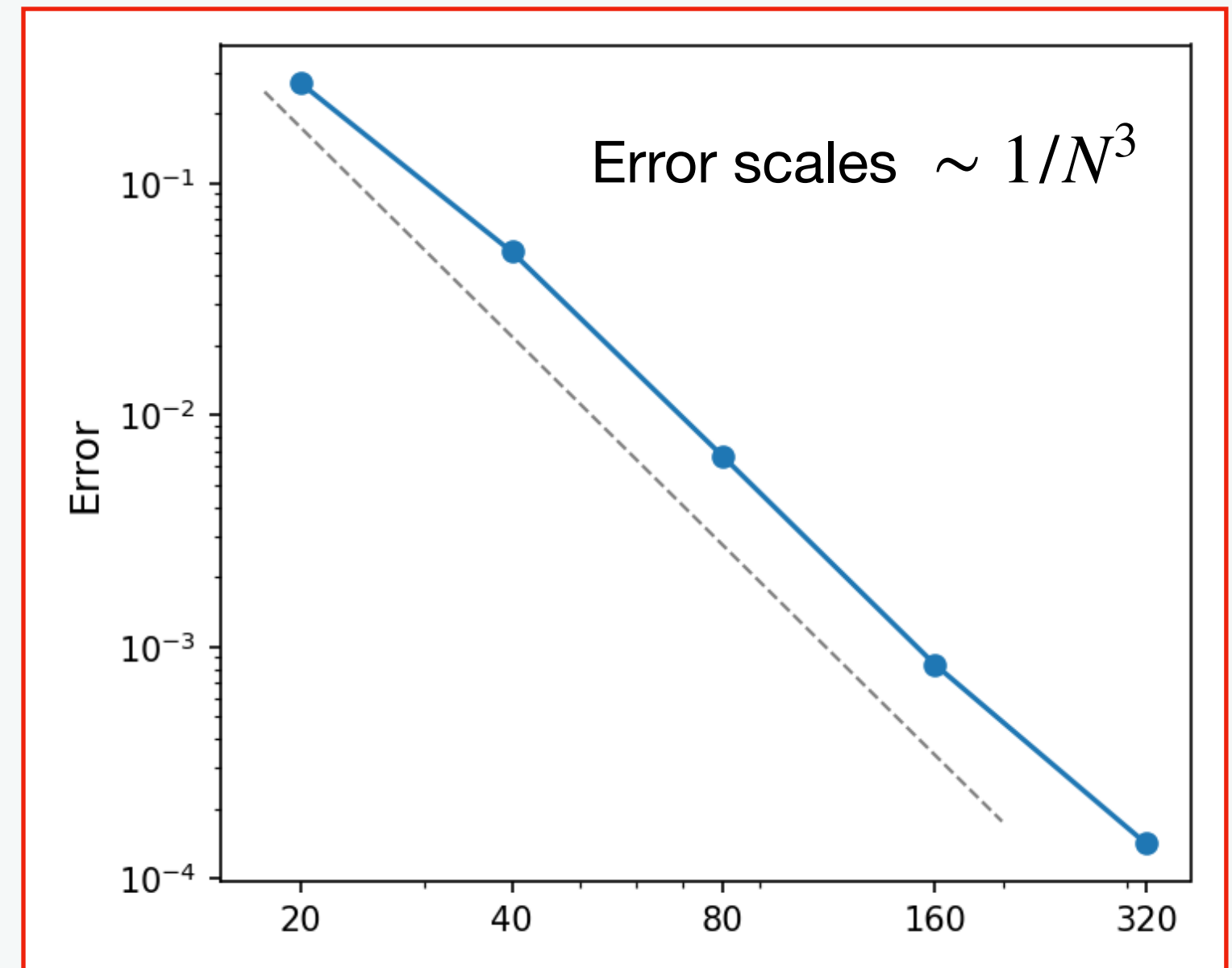
Vu+2022

e.g. Advection equation $\frac{\partial u}{\partial t} + \frac{\partial u}{\partial x} = 0$

analytic solution : $u(x, t) = u_0(x - t)$

Solve numerically with:

$$u_i(t_{n+1}) = u_i - \left(\frac{u_{i+1} - u_{i-1}}{2\Delta x} \right) \Delta t$$



Convergence of numerical methods

n th-order accurate scheme : truncation error scales as h^{n+1} or $(N_{\text{grid}})^{-(n+1)}$

Suppose we double (2x) the number of grid points per dimension, error drops...

- 1st order scheme : 1/4
- 2nd order scheme : 1/8
- ...

* spectral (or exponential) accuracy : h^N or $\exp(-N_{\text{grid}})$