

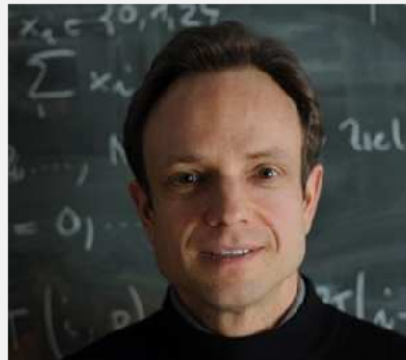
## Course Video

## Section 6.3.2: (Asymptotic) Convergence of Single-Step Methods

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(C) Seminar für Angewandte Mathematik, ETH Zürich

**Dependencies.** [Lecture → Section 3.2.2], [Lecture → Section 6.1.4], [Lecture → Section 6.3.1]

Duration:     minutes



Video and accompanying tablet notes may not match completely!

[Corrections and updates may have been made in tablet notes.]

VI. Numerical Integration –  
Single-Step Methods

## 6.3 General Single-Step Methods

## 6.3.1 (Asymptotic) Convergence of SSMs

$$\text{IVP: } \dot{y} = f(y), \quad y(0) = y_0 \in \mathcal{D} \subset \mathbb{R}^N$$

$$\text{SSM: } y_{k+1} = \mathcal{D}^{h_k} y_k, \quad h_k := t_{k+1} - t_k$$

on mesh  $\mathcal{M} = \{0 = t_0 < t_1 < \dots < t_m = T_0\}$

Focus: *h-convergence*, consider family of temporal meshes  
with  $h_m := \max_k h_k \xrightarrow{\quad} 0$

Discretization error:

$$\begin{aligned} & \max_{k=1}^m \|y_k - y(t_k)\| \quad [t \mapsto y(t) \stackrel{?}{=} \text{solution}] \\ & \|y_m - y(T)\| \end{aligned}$$

(2)

## Exp. 6.3.2.5 : (Speed of convergence of polygonal methods)

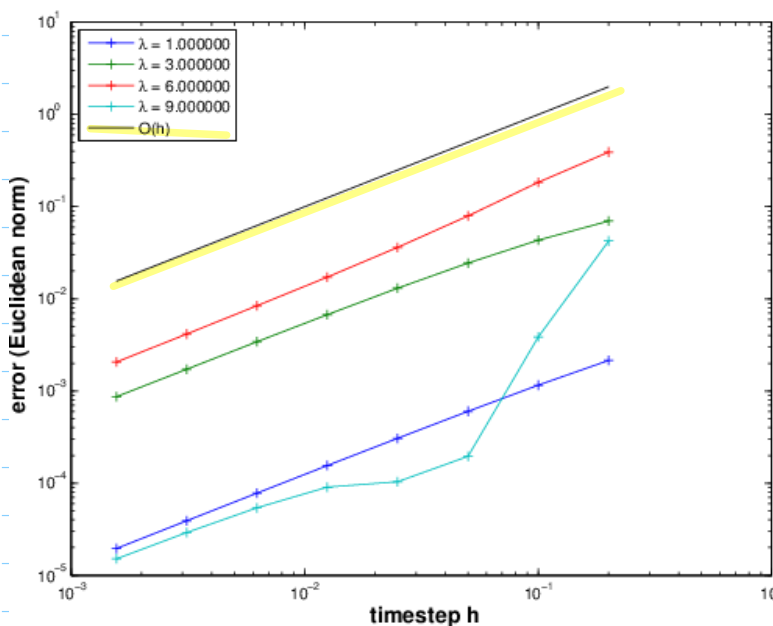
◆ We consider the following IVP for the **logistic ODE**, see Ex. 6.1.2.1

$$\dot{y} = \lambda y(1 - y) \quad , \quad y(0) = 0.01 \quad .$$

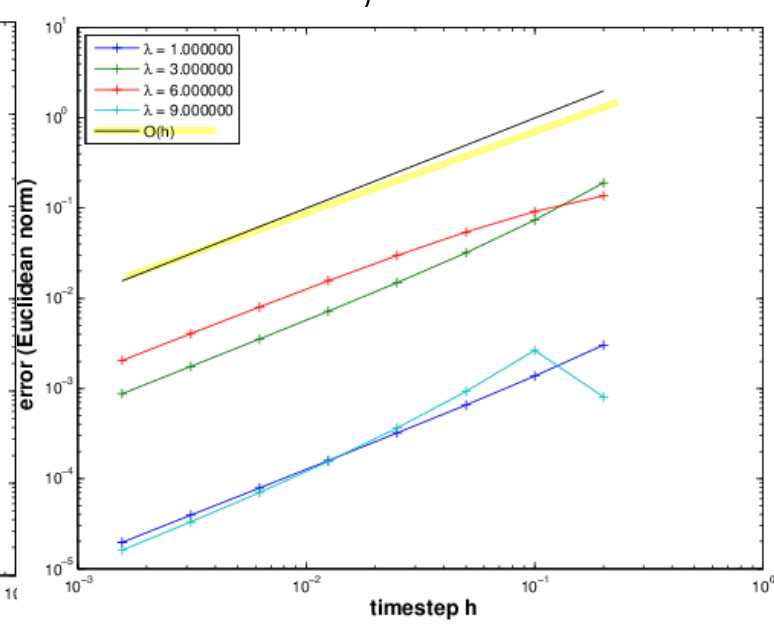
◆ We apply explicit and implicit Euler methods (6.2.1.4)/(6.2.2.2) with **uniform timestep**  $h = 1/M$ ,  $M \in \{5, 10, 20, 40, 80, 160, 320, 640\}$ .

◆ Monitored: Error at final time  $E(h) := |y(1) - y_M|$

equidistant  $\mathcal{M}$

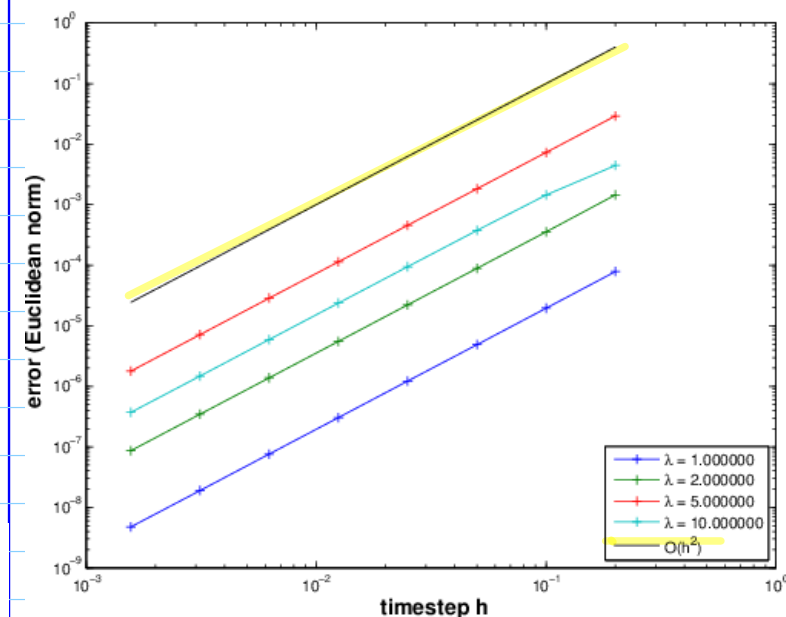


explicit Euler method



implicit Euler method

▷ Empiric **algebraic conv.** :  $E(h) = O(h) = O(M^{-1})$   
 [ rate/order = 1 ] for  $h \rightarrow 0$ ,  $M \rightarrow \infty$



implicit midpoint method

▷  $E(h) = O(h^2)$   
 Alg. conv. , rate/order = 2

### Algebraic convergence of single step methods

Consider the numerical integration of an initial value problem

$$\dot{\mathbf{y}} = \mathbf{f}(t, \mathbf{y}) \quad , \quad \mathbf{y}(t_0) = \mathbf{y}_0 \quad , \quad (6.1.3.2)$$

with **sufficiently smooth** right hand side function  $\mathbf{f} : I \times D \rightarrow \mathbb{R}^N$ .

Then conventional single step methods ( $\rightarrow$  Def. 6.3.1.4) will enjoy **algebraic convergence in the meshwidth**, more precisely, see ,

there is a  $p \in \mathbb{N}$  such that the sequence  $(\mathbf{y}_k)_k$  generated by the single step method for  $\dot{\mathbf{y}} = \mathbf{f}(t, \mathbf{y})$  on a mesh  $\mathcal{M} := \{t_0 < t_1 < \dots < t_M = T\}$  satisfies

$$\max_k \|\mathbf{y}_k - \mathbf{y}(t_k)\| \leq Ch^p \quad \text{for} \quad h := \max_{k=1, \dots, M} |t_k - t_{k-1}| \rightarrow 0 \quad , \quad (6.3.2.7)$$

with  $C > 0$  independent of  $\mathcal{M}$

$\rightarrow$  usually unknown

### Definition 6.3.2.8. Order of a single step method

The maximal integer  $p \in \mathbb{N}$  for which (6.3.2.7) holds for a single step method when applied to an ODE with (sufficiently) smooth right hand side, is called the **order** of the method.

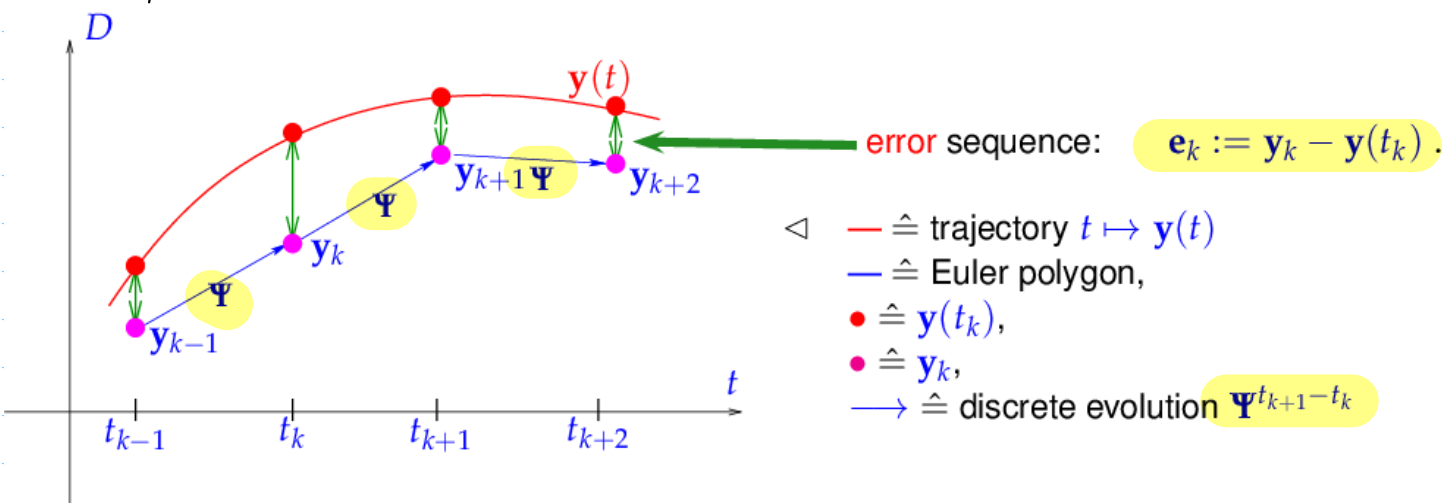
### § 6.3.2.9 Convergence analysis for the explicit Euler method

- IVP :  $\dot{y} = f(y)$  ,  $y(0) = y_0 \in D$
- Global Lipschitz condition : (L.C.)

$$\exists L > 0: \|f(y) - f(z)\| \leq L\|y - z\| \quad \forall y, z \in D, \quad (6.3.2.10)$$

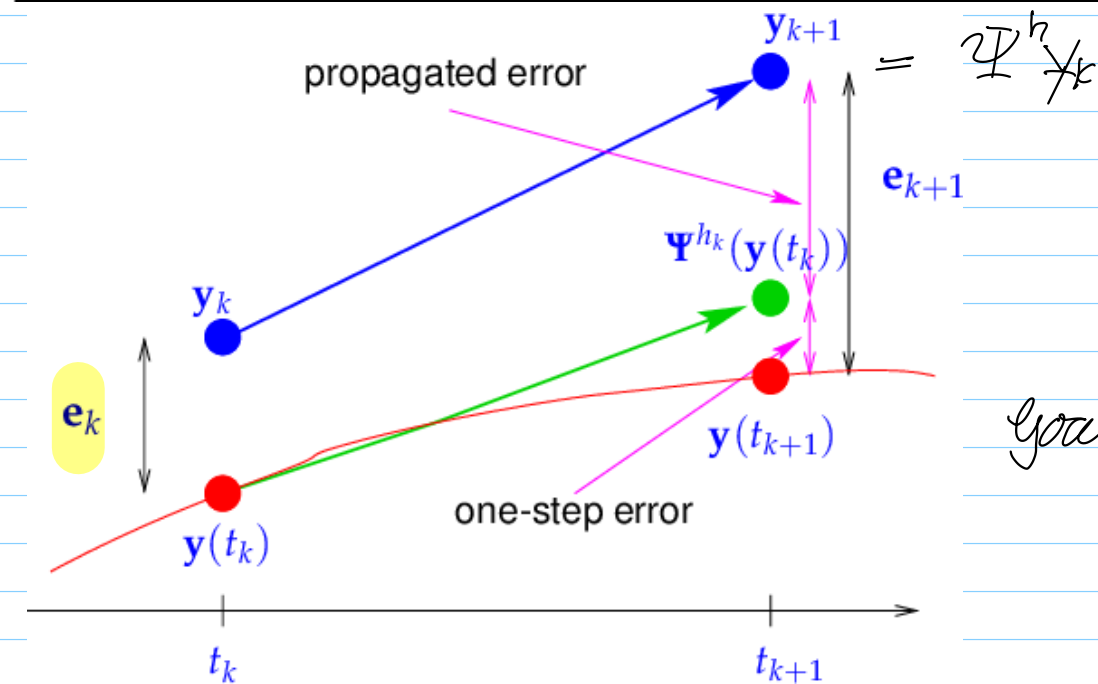
Expl. Euler:  $y_{k+1} = y_k + h_k f(y_k)$  ,  $k = 1, \dots, M-1$ .

(6.2.1.4)



### Fundamental error splitting:

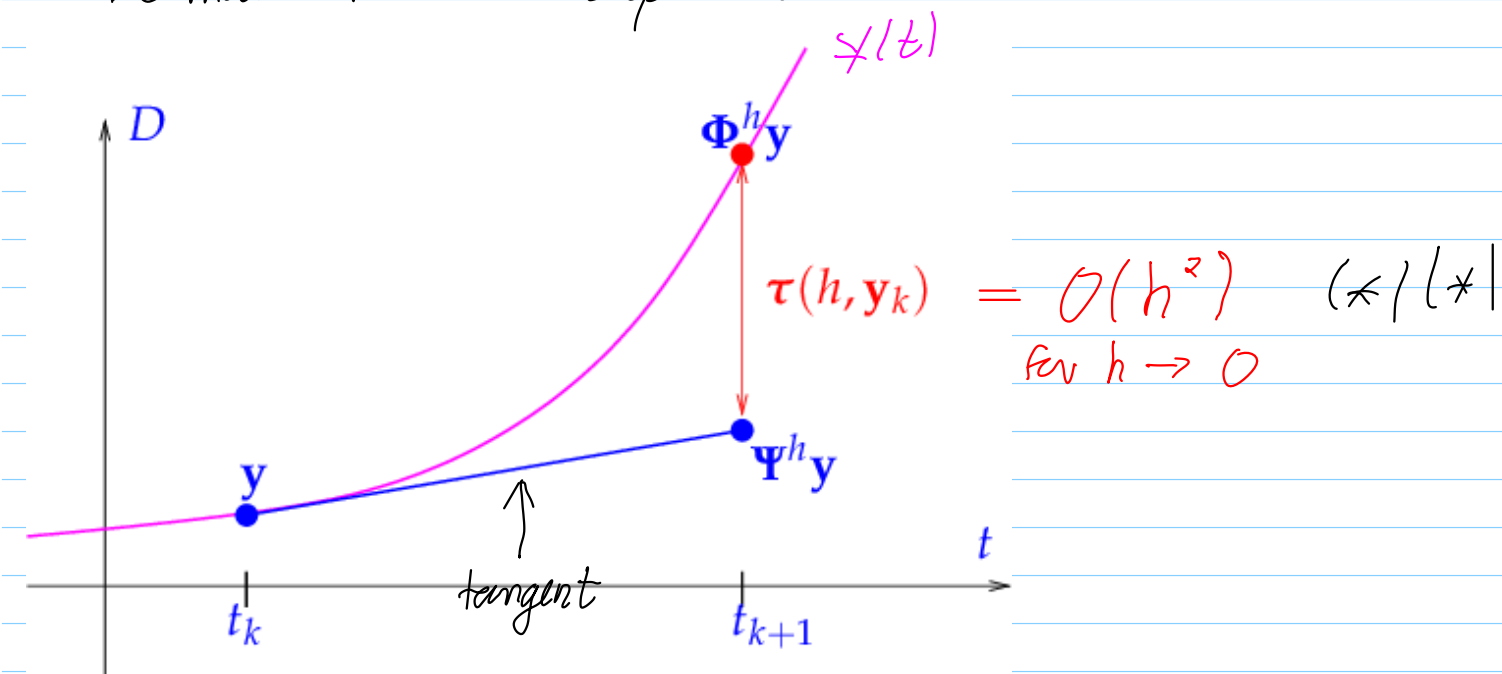
$$\begin{aligned} e_{k+1} &= \overbrace{\Psi^{h_k} y_k}^{y_{k+1}} - \overbrace{\Phi^{h_k} y(t_k)}^{y(t_{k+1})} \\ &= \underbrace{\Psi^{h_k} y_k - \Psi^{h_k} y(t_k)}_{\text{propagated error}} \\ &\quad + \underbrace{\Psi^{h_k} y(t_k) - \Phi^{h_k} y(t_k)}_{\text{one-step error}}. \end{aligned} \quad (6.3.2.12)$$



Goal: recursion for  $e_k$

4)

- Estimate for one-step error:



- Estimate for propagated error:

$$\underbrace{\|\Psi^h y_k - \Psi^h y(t_k)\| = \|y_k + hf(y_k) - y(t_k) - hf(y(t_k))\|}_{L.C.} \leq (1+hL) \|y_k - y(t_k)\|$$

+ (6.3.2.12)

$$[\varepsilon_k := \|e_k\|]: \quad \varepsilon_{k+1} \leq (1+hL) \varepsilon_k + \|\tau(h, y_k)\|$$

$$(*) \quad \varepsilon_{k+1} \leq (1+hL) \varepsilon_k + O(h^2) \Rightarrow \varepsilon_k = O(h)$$

Total error arises from accumulation of propagated one-step errors!

True for all consistent SSMs

### Order of algebraic convergence of single-step methods

Consider an IVP (6.1.3.2) with solution  $t \mapsto y(t)$  and a single step method defined by the discrete evolution  $\Psi$  ( $\rightarrow$  Def. 6.3.1.4). If the one-step error along the solution trajectory satisfies ( $\Phi$  is the evolution map associated with the ODE, see Def. 6.1.4.3)

$$\|\Psi^h y(t) - \Phi^h y(t)\| \leq Ch^{p+1} \quad \forall h \text{ sufficiently small, } t \in [0, T], \quad (6.3.2.22)$$

for some  $p \in \mathbb{N}$  and  $C > 0$ , then, usually,

$$\max_k \|y_k - y(t_k)\| \leq \bar{C} h_{\mathcal{M}}^p,$$

with  $\bar{C} > 0$  independent of the temporal mesh  $\mathcal{M}$ : The (pointwise) discretization error converges algebraically with order/rate  $p$ .

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## Ex. 6.3.2.25 (Order of finite-difference SSMs)

Implicit midpoint method (IMM)

$$\Psi^\tau y := w(\tau) : w(\tau) = y + \tau f\left(\frac{1}{2}(y + w(\tau))\right)$$

Tool: Taylor expansions

$$\begin{aligned}\Phi^\tau y &= y + \frac{\partial \phi}{\partial t}(0, y) \tau + \frac{1}{2} \frac{\partial^2 \phi}{\partial t^2}(0, y) \tau^2 + O(\tau^3) \\ &= y + f(y) \tau + \frac{1}{2} \tau^2 Df(y) f(y) + O(\tau^3)\end{aligned}$$

↑  
by chain rule

$$w(\tau) = y + \tau f\left(y + \frac{1}{2} \tau f\left(\frac{1}{2}(y + w(\tau))\right)\right)$$

Then Taylor expansion of  $f$  around  $y$ 

$$\triangleright w(\tau) = y + \tau f(y) + \frac{1}{2} \tau^2 Df(y) f(y) + O(\tau^3)$$

$$\triangleright \Phi^\tau y - \Psi^\tau y = O(\tau^3)$$

IMM has order 2

## Please answer review questions

After you have watched the video take a short break and then try to answer the

## review questions 6.3.2.33

Please do not flip pages of the lecture document, nor look at tablet notes.



