

Numerical Solutions to ODEs

CMPT 882

Jan. 23

Numerical Solutions of ODEs

- In general, $\dot{x} = f(x, u)$ does not have a closed-form solution
 - Instead, we usually compute numerical approximations to simulate system behaviour
 - Done through discretization: $t^k = kh$, $u^k := u(t^k)$
 - h represents size of time step
 - Goal: compute $y^k \approx x(t^k)$
- Key considerations
 - Consistency: Does the approximation satisfy the ODE as $h \rightarrow 0$?
 - Accuracy: How fast does the solution converge?
 - Stability: Do approximation error remain bounded over time?
 - Convergence: Does the solution converge the true solution as $h \rightarrow 0$?

Euler Methods

- ODE: $\dot{x} = f(x, u)$
 - Discretization: $t^k = kh$, $u^k := u(t^k)$
 - Want: Approximate solution: $y^k \approx x(kh)$

$$\begin{aligned} \dot{x} &= f(x, u) \\ \frac{x(t^{k+1}) - x(t^k)}{h} &\approx f(x(t^k), u^k) \\ \frac{y^{k+1} - y^k}{h} &= f(y^k, u^k) \end{aligned}$$

- Forward Euler
 - Most naïve method (explicit method)

$$\frac{y^{k+1} - y^k}{h} = f(y^k, u^k) \Rightarrow y^{k+1} = y^k + hf(y^k, u^k)$$

- Backward Euler

$$\frac{y^{k+1} - y^k}{h} = f(y^{k+1}, u^k) \Rightarrow \text{solve for } y^{k+1} \text{ implicitly}$$

Visualizing Euler Methods

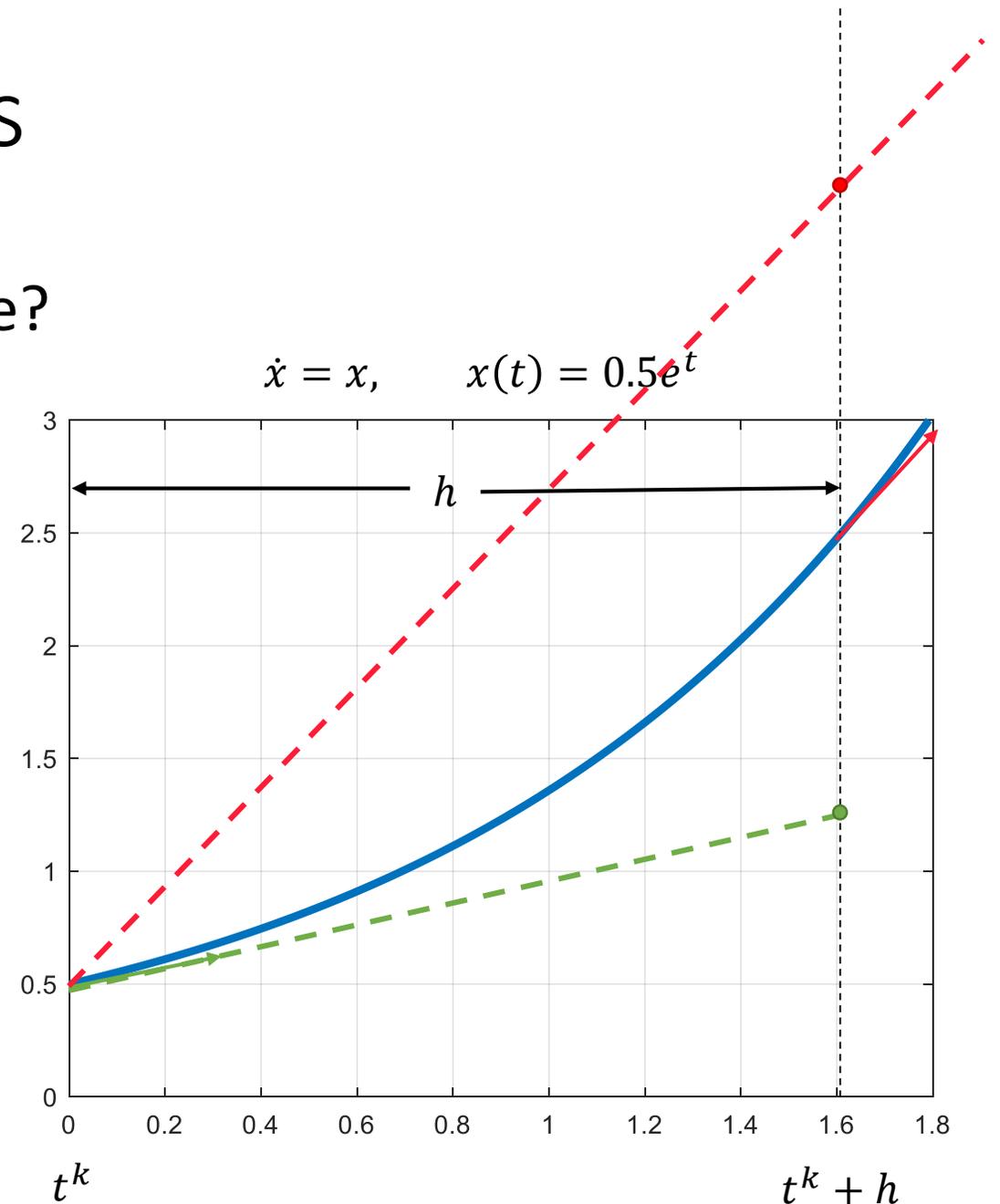
- Main consideration: what slope to use?

- Forward Euler: slope at beginning

$$y^{k+1} = y^k + hf(y^k, u^k)$$

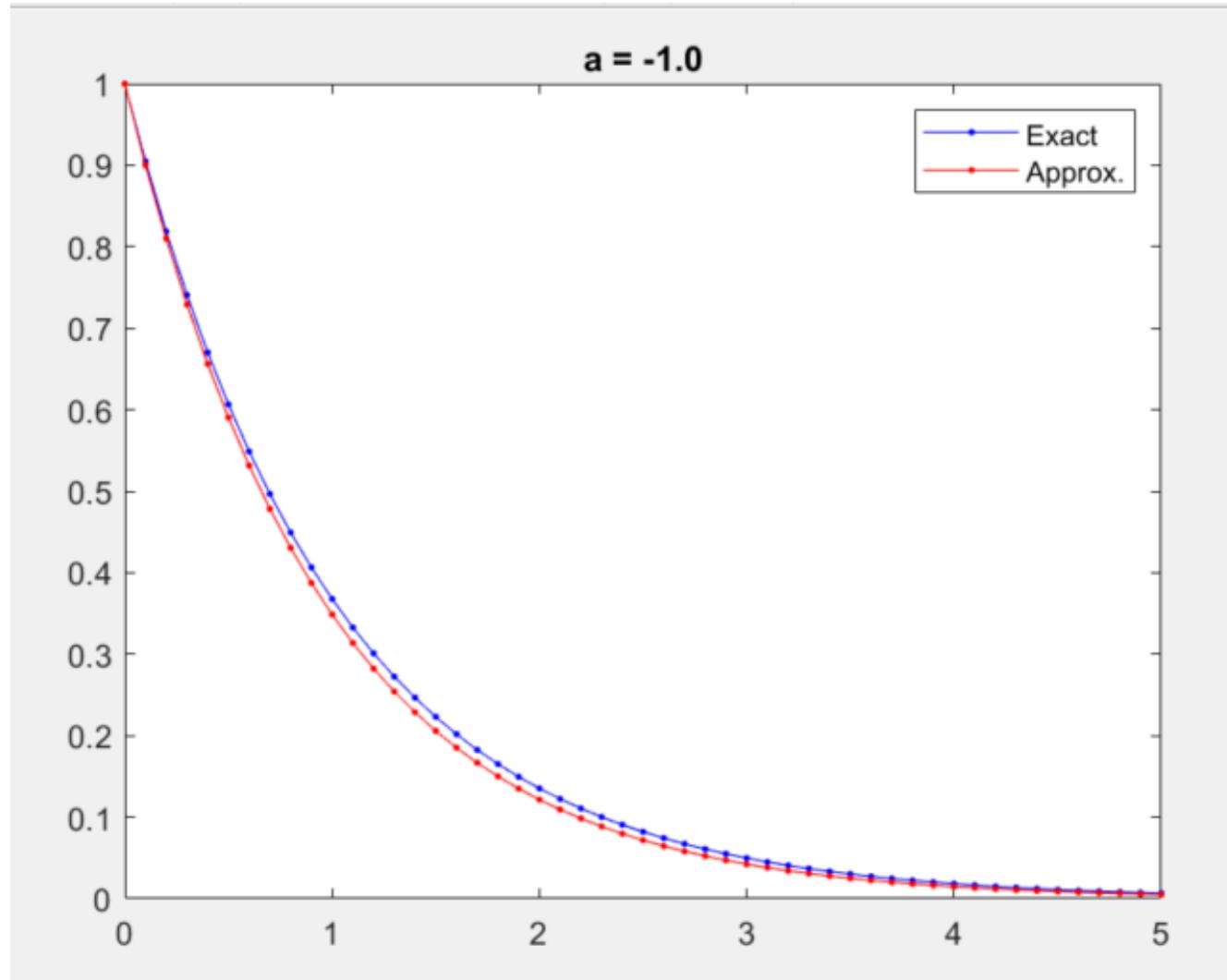
- Backward Euler: slope at the end

$$y^{k+1} = y^k + hf(y^{k+1}, u^k)$$



Example

- $\dot{x} = ax, x(0) = x_0$
 - Analytic solution: $x(t) = x_0 e^{at}$
- Forward Euler
 - $y^{k+1} = y^k + hf(y^k, u^k)$
 - $y^{k+1} = y^k + hay^k$
 - $y^{k+1} = (1 + ha)y^k$



Example

```
%% Problem setup
```

```
x0 = 1;
```

```
a = -1;
```

```
h = 0.1;
```

```
T = 5;
```

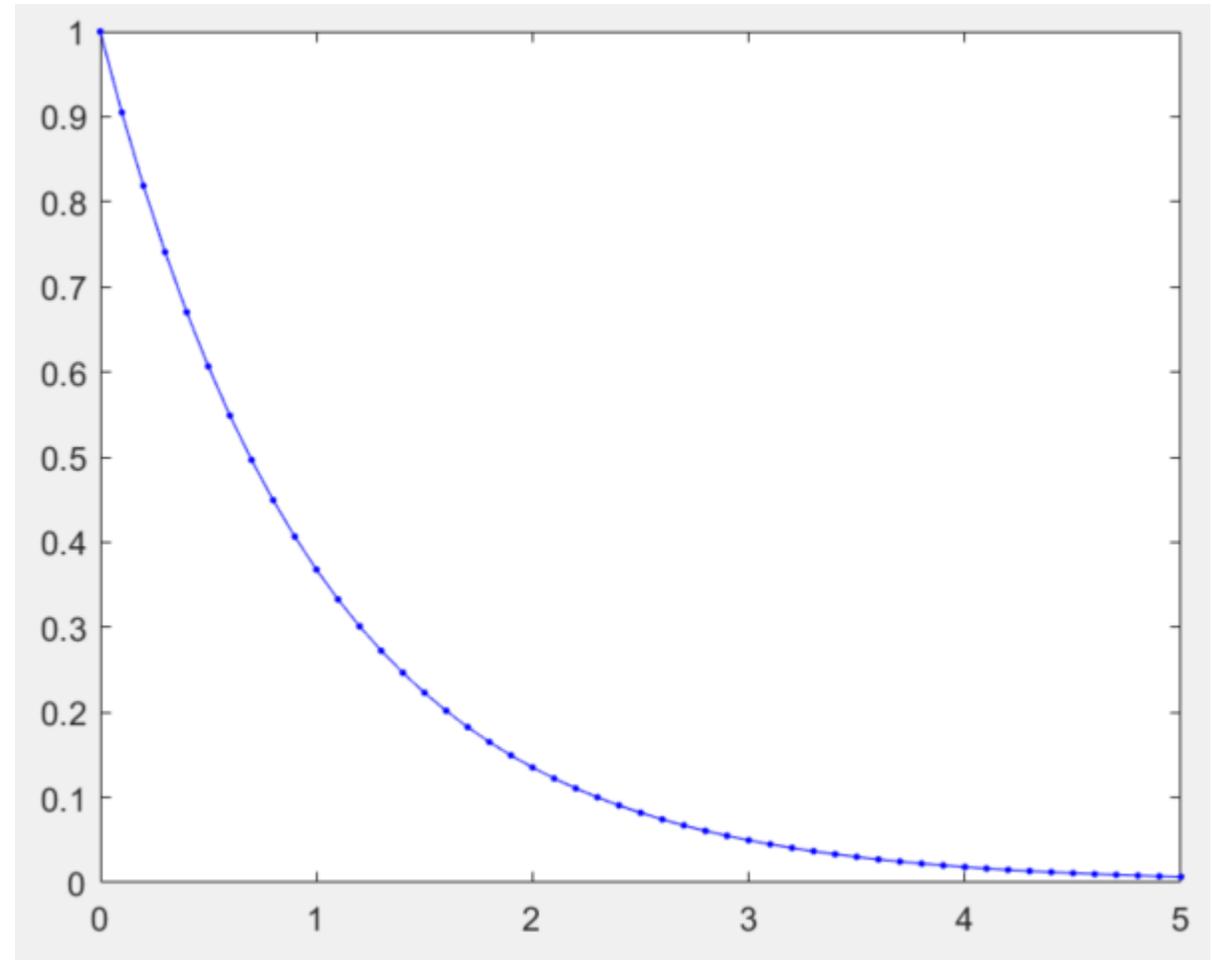
```
tau = 0:h:T;
```

```
%% Exact solution
```

```
x_exact = @(t) exp(a*t);
```

```
figure
```

```
plot(tau, x_exact(tau), 'b.-')
```

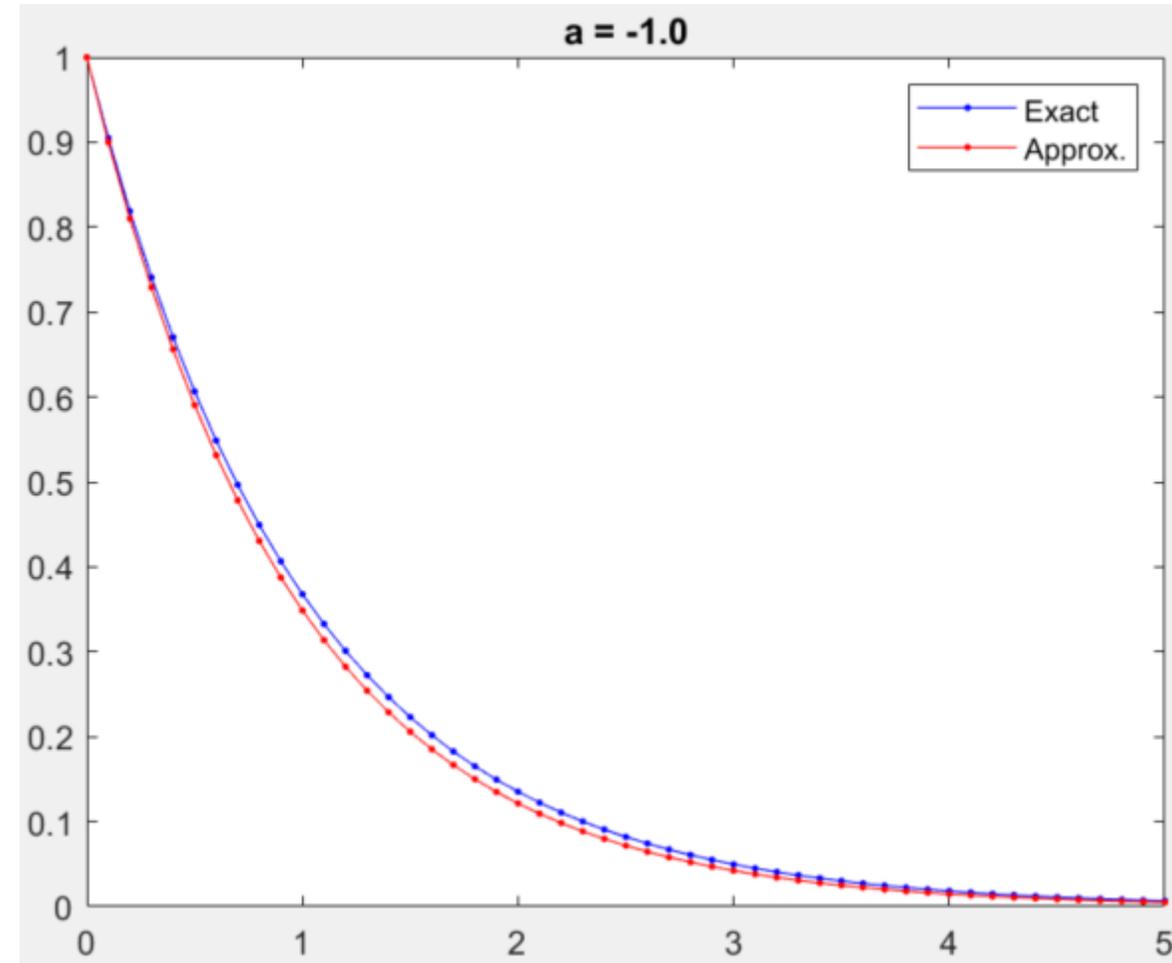


Example

```
%% Forward Euler
f = @(x) a*x;
y_approx = -ones(size(tau));
y_approx(1) = x0;

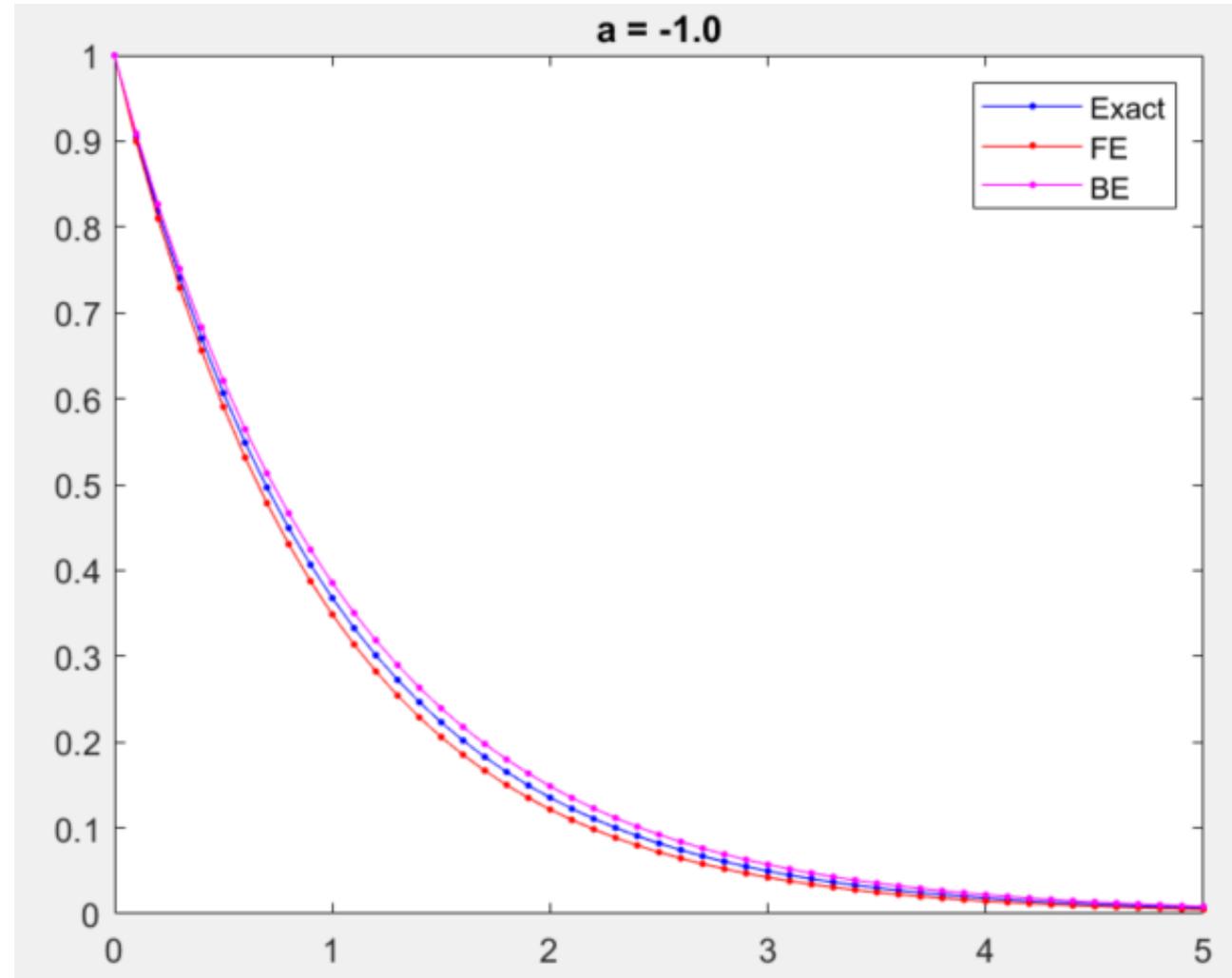
% Initialize vector
for i = 2:length(tau)
    y_approx(i) = y_approx(i-1)*(1+h*a);
end

hold on
plot(tau, y_approx, 'r.-')
title(sprintf('a = %.1f', a))
legend('Exact', 'Approx.')
```



Example

- $\dot{x} = ax, x(0) = x_0$
 - Analytic solution: $x(t) = x_0 e^{at}$
- Backward Euler
 - $y^{k+1} = y^k + hf(y^{k+1})$
 - $y^{k+1} = y^k + hay^{k+1}$
 - $y^{k+1} - hay^{k+1} = y^k$
 - $(1 - ha)y^{k+1} = y^k$
 - $y^{k+1} = \frac{y^k}{1-ha}$



Numerical Consistency: Forward Euler

- **Consistency:** ODE is satisfied as $h \rightarrow 0$

- Forward Euler: $y^{k+1} = y^k + hf(y^k, u^k)$ $\frac{y^{k+1} - y^k}{h} = f(y^k, u^k)$

- **Local truncation error:** Analysis requires $\frac{\|e^k\|}{h} \rightarrow 0$ as $h \rightarrow 0$

- $\|e^k\|$: Error induced during one step, assuming perfect previous information
- Forward Euler approximate solution:

$$y^{k+1} = x(t^k) + hf(x(t^k), u^k)$$

- True solution:

$$\begin{aligned} x(t^{k+1}) &= x(t^k + h) = x(t^k) + h \frac{dx}{dt}(t^k) + \frac{h^2}{2} \frac{d^2x}{dx^2}(t^k) + O(h^3) \\ &= x(t^k) + hf(x(t^k), u^k) + \frac{h^2}{2} \frac{d^2x}{dx^2}(t^k) + O(h^3) \end{aligned}$$

Numerical Consistency: Forward Euler

- Local truncation error:
$$\begin{aligned} e^k &= x(t^{k+1}) - y^{k+1} \\ &= x(t^k) + hf(x(t^k), u^k) + \frac{h^2}{2} \frac{d^2x}{dx^2}(t^k) + O(h^3) - \left(x(t^k) + hf(x(t^k), u^k) \right) \\ &= \frac{h^2}{2} \frac{d^2x}{dx^2}(t^k) + O(h^3) \\ &= O(h^2) \end{aligned}$$
- Consistency requires $\frac{\|e^k\|}{h} \rightarrow 0$ as $h \rightarrow 0$
$$\frac{\|e^k\|}{h} = \frac{\left| \frac{h^2}{2} \frac{d^2x}{dx^2}(t^k) + O(h^3) \right|}{h} = \left| \frac{h}{2} \frac{d^2x}{dx^2}(t^k) + O(h^2) \right| \rightarrow 0$$
- If $\frac{\|e^k\|}{h} = O(h^p)$, then the numerical method is “order p ”.
 - Forward Euler is an order 1 method, or first order method

Numerical Consistency

- More generally: $y^{k+1} = \sum_{n=k_1}^k \alpha_n y^n + h \sum_{n=k_2}^k \beta_n f(y^n, u^n)$

- Truncation error:

$$e^k := x(t^{k+1}) - \sum_{n=k_1}^k \alpha_n x(nh) - h \sum_{n=k_2}^k \beta_n f(x(nh), u^n)$$

- Consistency requires $\frac{\|e^k\|}{h} \rightarrow 0$ as $h \rightarrow 0$

- If $\frac{\|e^k\|}{h} = O(h^p)$, then the numerical method is “order p ”.

Numerical Stability: Forward Euler

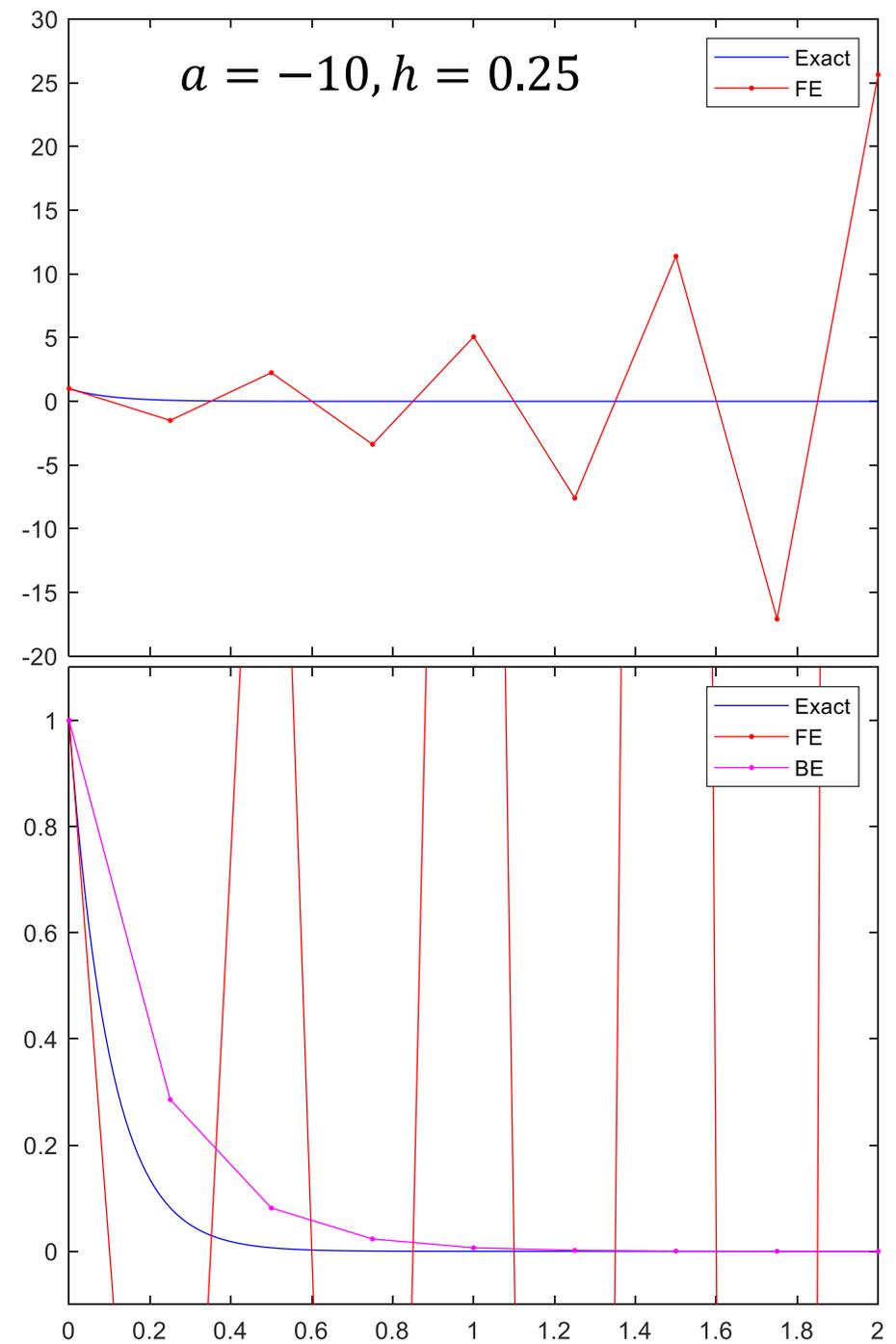
- $y^{k+1} = y^k + hf(y^k, u^k)$
 - A map from y^k to y^{k+1}
 - Stability means y^k does not “blow up” when the true solution $x(t^k)$ is bounded
 - Usually, stability requires that the time step h cannot be too large
- Example: $\dot{x} = ax, a < 0$
 - $y^{k+1} = (1 + ah)y^k$
 - Stability requires $|1 + ah| \leq 1 \Leftrightarrow -ah \leq 2$
 - For $a = -10$, we have $|1 - 10h| \leq 1 \Leftrightarrow h \leq 0.2$

Numerical Stability: Backward Euler

- $y^{k+1} = y^k + hf(y^{k+1}, u^k)$
 - A map from y^k to y^{k+1}
 - Stability means y^k does not “blow up” when the true solution $x(t^k)$ is bounded
 - Usually, stability requires that the time step h cannot be too large
- Example: $\dot{x} = ax, a < 0$
 - $y^{k+1} = \frac{y^k}{1+ah}$
 - Stability requires $\left| \frac{1}{1-ah} \right| \leq 1$
 - No restrictions on h , for any a !

Numerical Stability

- Example: $\dot{x} = ax$ with forward Euler
 - If $a = -10$, $h \leq 0.2$ is required for stability
- Example 2: $\dot{x} = ax$ with backward Euler
 - No restrictions on h , for any a



Numerical Stability

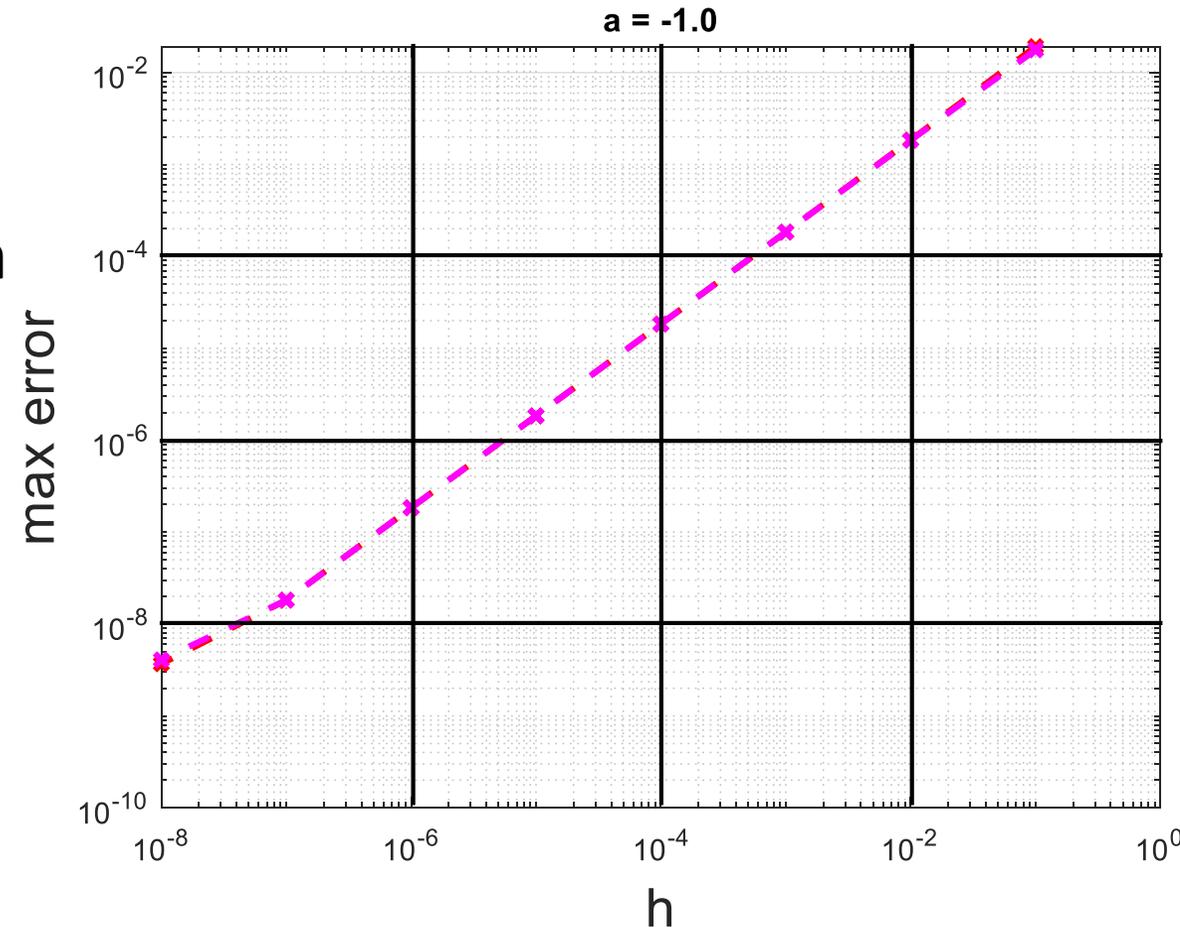
- More generally: $y^{k+1} = \sum_{n=k_1}^k \alpha_n y^n + h \sum_{n=k_2}^k \beta_n f(y^n, u^n)$
 - Desired property: the approximation y^k does not “blow up” when the true solution $x(t^k)$ is bounded
 - Usually, this means time step h cannot be too large
- Specifically, one typically considers $\dot{x} = ax, a < 0$.
 - A stable numerical approximation to $\dot{x} = ax, a < 0$ has the property that $y^k \rightarrow 0$

Numerical Convergence

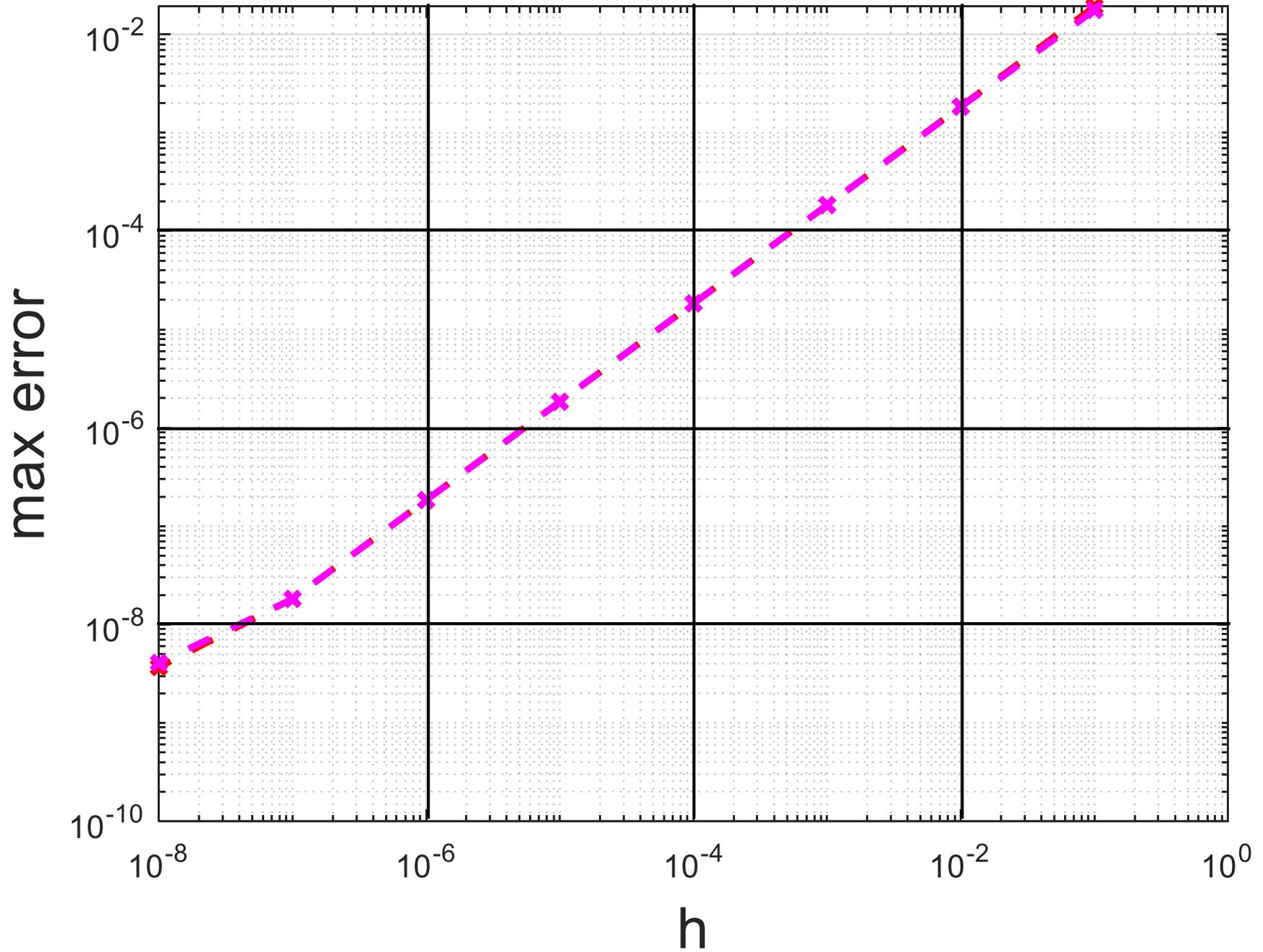
- **Convergence:** $\max_k \|x(t^k) - y^k\| \rightarrow 0$ as $h \rightarrow 0$
 - Maximum error goes to zero as time step goes to 0
- Dahlquist Equivalence Theorem
 - Consistency + stability \Leftrightarrow convergence
- Convergence rate
 - For order p methods: $\max_k \|x(t^k) - y^k\| \leq O(h^{p+1})$
 - Forward and backward Euler: $p = 1$
 - If we half h , then the error also halves

Numerical Convergence

- Visualize convergence rate with Max error vs. h plot
- Forward and backward Euler are both 1st order
 - Half the size of h leads to half the error
- Usually, log-log plots are used to show a wide range of errors and h
 - Order p method has a slope of p (approximately).



a = -1.0



Stiff equations

- ODEs with components that have very fast rates of change
 - Usually requires very small step sizes for stability
- Example: $\dot{x}_1 = ax_1$ with forward Euler
 - Stability requires $|1 + ha| \leq 1$
 - For $a = -100$, we have $|1 - 100h| \leq 1 \Leftrightarrow h \leq 0.02$
- Small step size is required even if there are other slower changing components like $\dot{x}_2 = x_1 - x_2$
 - Implicit methods (eg. backward Euler) are useful here

$$\begin{aligned}\dot{x}_1 &= -100x_1 \\ \dot{x}_2 &= x_1 - x_2\end{aligned}$$

$$\dot{x} = \begin{bmatrix} -100 & 0 \\ 1 & -1 \end{bmatrix} x$$