# Numerical Analysis & Scientific Computing II

# Module 2 Initial Value Problems

- 2.1 Well-posedness
- 2.2 Stability
- 2.3 Euler's method
  - Derivations





**Derivation using Taylor series** 



## **Derivation using Taylor series**

Consider the Taylor series

$$y(t_{k+1}) = y(t_k) + (t_{k+1} - t_k)y'(t_k) + \frac{(t_{k+1} - t_k)^2}{2}y''(t_k) + \cdots$$



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## Derivation using finite difference approximation

Replacing the y'(t) in the ODE y' = f(t, y) by a first order forward difference approximation, we obtain an algebraic equation

$$\frac{y_{k+1} - y_k}{t_{k+1} - t_k} = f(t_k, y_k)$$

that yields the Euler's method.



Derivation using polynomial interpolation



## **Derivation using polynomial interpolation**

One point Hermite polynomial p(t) that matches the function and derivative data at  $t = t_k$ , that is,

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$$y_{k+1} = \alpha y_k + \beta y_k'$$

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  $t_{k+1}e = \alpha t_k e + \beta e$ 



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implying  $\alpha = 1$  and  $\beta = t_{k+1} - t_k$  resulting in the Euler's method.

# Numerical Analysis & Scientific Computing II

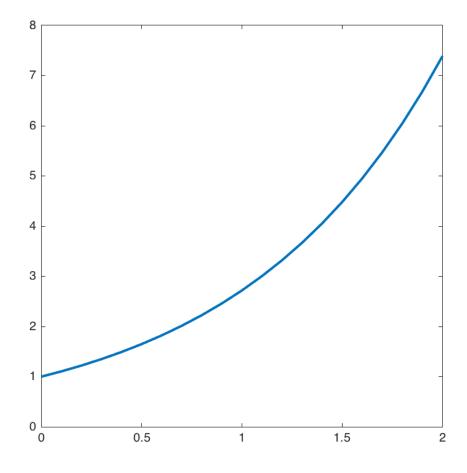
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- 2.1 Well-posedness
- 2.2 Stability
- 2.3 Euler's method
  - Errors and error propagation



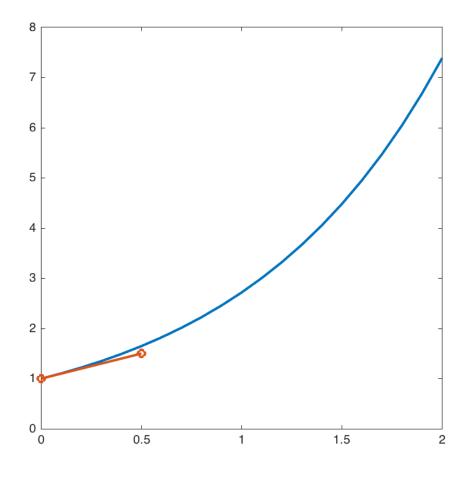


## **Example**



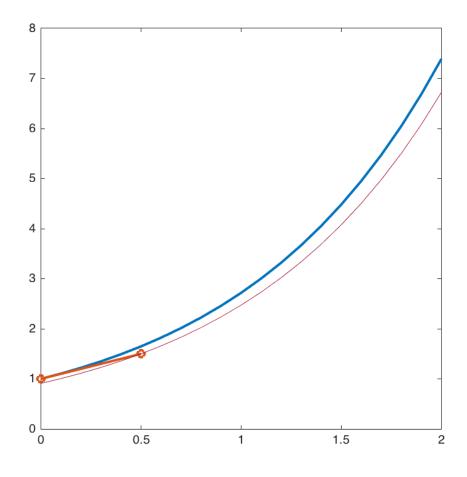


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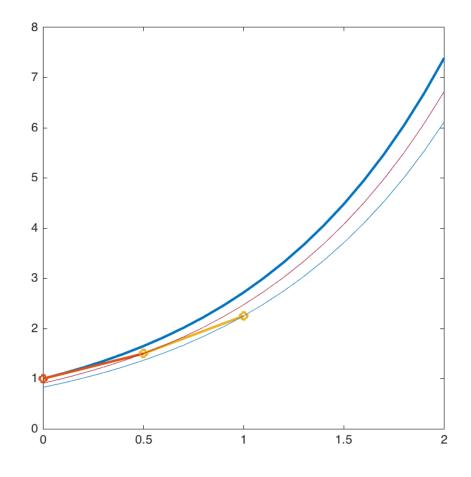


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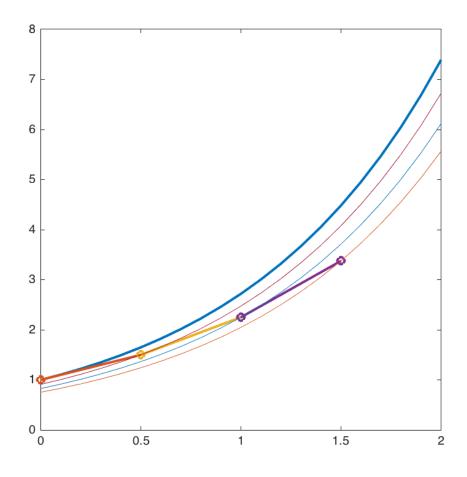


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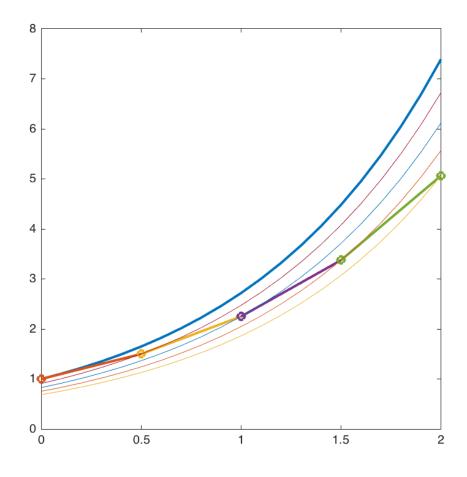


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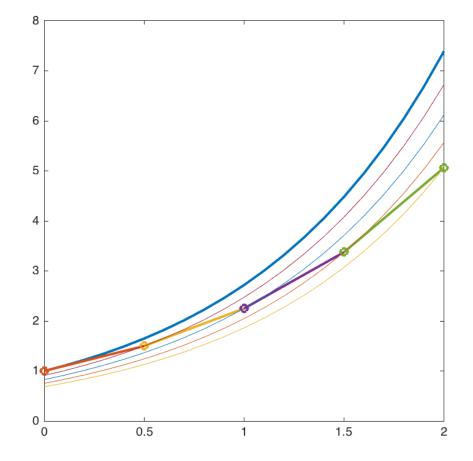




## **Example**

Let us solve y' = y, y(0) = 1 using the Euler's method taking the uniform step size  $h = h_k = t_{k+1} - t_k = 0.5$ .  $y_1 = 1 + h$ ,  $y_2 = y_1 + hy_1 = (1 + h)^2$ ,  $y_3 = y_2 + hy_2 = (1 + h)^3$ , ...,  $y_k = (1 + h)^k$ , ...

An error is introduced at each step of the method – one step error or single step error.



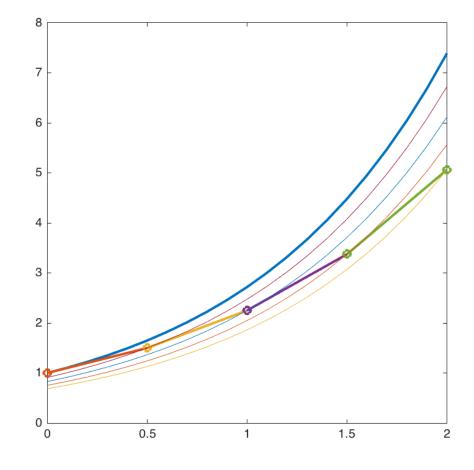


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These local errors accumulate over time to produce a global error.





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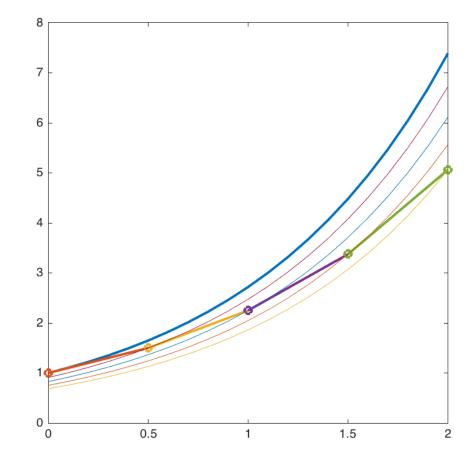
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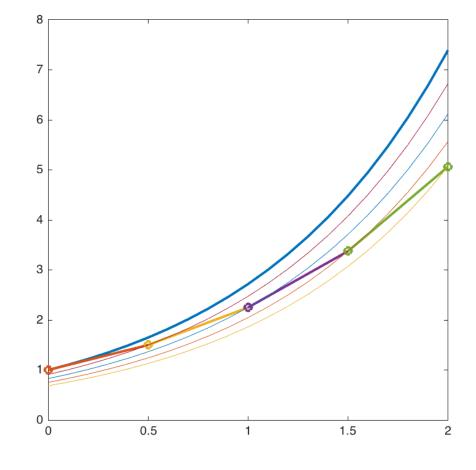
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With each step, we hop from one solution of the ODE to another.



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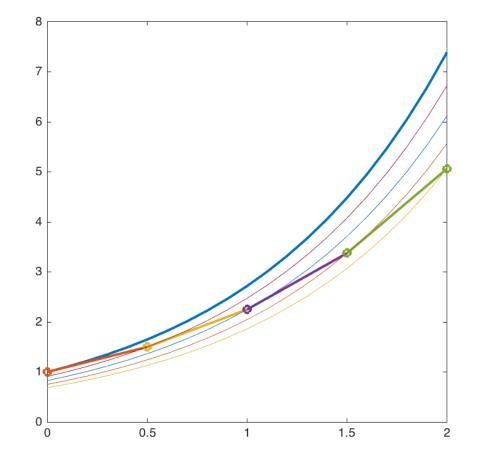
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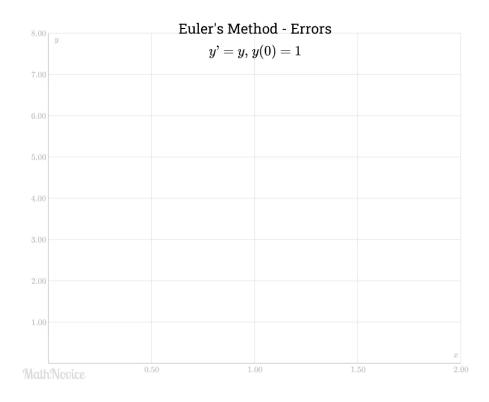
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Source: https://youtu.be/714HgSO0h7g



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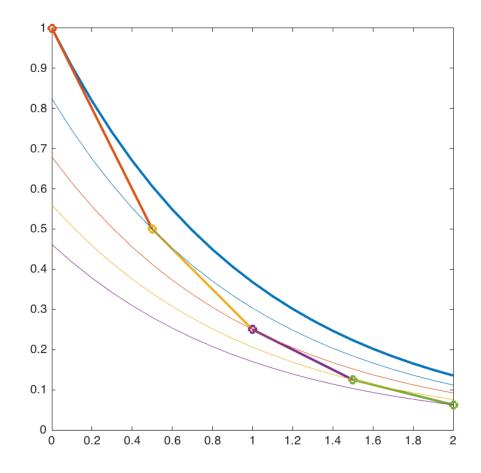
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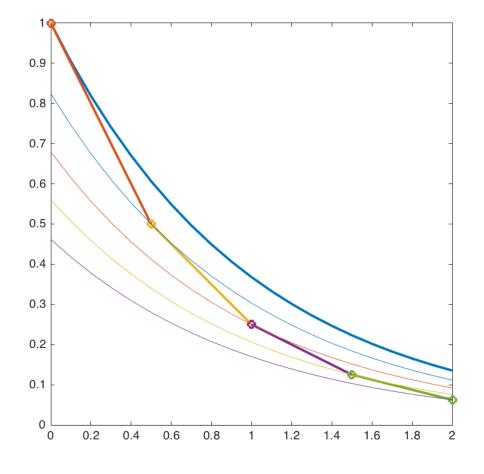




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For an equation with stable solutions, the errors in the numerical solution do not grow, and for equations with asymptotically stable solutions, the errors diminish with time.

## Initial Value Problems: Accuracy and Stability



## Sources of error

## Rounding error –

... due to truncation of data (e.g., real numbers requiring infinite space is represented using a finite amount of space); finite precision of the floating point arithmetic.

#### Truncation error -

... due to truncation of infinite mathematical processes to finite processes or algorithms (e.g., derivative replaced by a finite difference approximation, or integration replaced by a quadrature).

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In most practical situations, truncation error dominates other sources and, therefore, in analysis of numerical methods, we will focus exclusicely on truncation error.