# Numerical Methods for Differential Equations <br> Course objectives and preliminaries 

Tony Stillfjord, Gustaf Söderlind

Numerical Analysis, Lund University


1. Course structure and objectives
2. What is Numerical Analysis?
3. The four principles of Numerical Analysis

## 1. Course structure and objectives Learning by doing

Main objectives

- Learn basic scientific computing for solving differential equations
- Understand mathematics-numerics interaction, and how to match numerical method to mathematical properties
- Understand correspondence between principles in physics and mathematical equations
- Construct and use elementary Matlab/Python programs for differential equations


## A motto

The computer is to mathematics what the telescope is to astronomy, and the microscope is to biology

- Peter D Lax

The operators we will deal with

Initial value problems

$$
\frac{\mathrm{d}}{\mathrm{~d} t}
$$

The operators we will deal with

## Boundary value problems

$$
\frac{d^{2}}{d x^{2}}
$$



The operators we will deal with

## Partial differential operators

## Parabolic

## Hyperbolic

$$
\frac{\partial}{\partial t}-\frac{\partial^{2}}{\partial x^{2}}
$$



## Course topics

- IVP Initial value problems first-order equations; higher-order equations; systems of differential equations
- BVP Boundary value problems two-point boundary value problems; Sturm-Liouville eigenvalue problems
- PDE Partial differential equations diffusion equation; advection equation; convection-diffusion; wave equation
- Applications in all three areas


## Initial value problems

A first-order equation without elementary analytical solution

$$
\frac{\mathrm{d} y}{\mathrm{~d} t}=y-\mathrm{e}^{-t^{2}}, \quad y(0)=y_{0}
$$

A second-order equation Motion of a pendulum

$$
\ddot{\theta}+\frac{g}{L} \sin \theta=0, \quad \theta(0)=\theta_{0}, \quad \dot{\theta}(0)=\dot{\theta}_{0}
$$

A system of equations Predator-prey model

$$
\begin{aligned}
& \dot{y}_{1}=k_{1} y_{1}-k_{2} y_{1} y_{2} \\
& \dot{y}_{2}=k_{3} y_{1} y_{2}-k_{4} y_{2}
\end{aligned}
$$

where $y_{1}$ is the prey population and $y_{2}$ is the predator species

## Boundary value problems

Second-order two-point BVP Electrostatic potential u between concentric metal spheres

$$
\frac{\mathrm{d}^{2} u}{\mathrm{~d} r^{2}}+\frac{2}{r} \frac{\mathrm{~d} u}{\mathrm{~d} r}=0, \quad u\left(R_{1}\right)=V_{1}, \quad u\left(R_{2}\right)=0
$$

at distance $r$ from the center

A Sturm-Liouville eigenvalue problem Euler buckling of a slender column

$$
y^{\prime \prime}=\lambda y, \quad y^{\prime}(0)=0, \quad y(1)=0
$$

Find eigenvalues $\lambda$ and eigenfunctions $y$

## Some application areas in ODEs

Initial value problems

- mechanics
- electrical circuits

$$
\begin{aligned}
& M \ddot{q}=F(q) \\
& C \dot{v}=-I(v) \\
& \dot{c}=f(c)
\end{aligned}
$$

Boundary value problems

- materials
- microphysics
$u^{\prime \prime}=M / E I$
- eigenmodes

$$
-\frac{\hbar}{2 m} \psi^{\prime \prime}=E \psi
$$

$-u^{\prime \prime}=\lambda u$

## Partial differential equations

The Poisson Equation

$$
u_{x x}+u_{y y}=f(x, y)
$$

subject to

$$
u(x, y)=g(x, y)
$$

on the boundary $x=0 \vee x=1$, and $y=0 \vee y=1$ is an elliptic PDE modelling e.g. the displacement $u$ of an elastic membrane under load $f$

## Partial differential equations

The Diffusion Equation

$$
u_{t}=u_{x x}+f(x)
$$

subject to

$$
\begin{aligned}
& u(x, 0)=g(x), \quad x \in[0, L] \\
& u(0, t)=c_{1}, \quad u(L, t)=c_{2}, \quad t \in[0, T]
\end{aligned}
$$

is a parabolic PDE modelling e.g. temperature $u$ as a funtion of time, in a rod with temperature $c_{1}$ and $c_{2}$ at its ends, with initial temperature distribution $g(x)$ and source term $f(x)$

## Partial differential equations

The Wave Equation

$$
u_{t t}=u_{x x}
$$

subject to

$$
\begin{aligned}
& u(x, 0)=g(x), \quad x \in[0, L] \\
& u(0, t)=0, \quad u(L, t)=0, \quad t \in[0, T]
\end{aligned}
$$

is a hyperbolic PDE modelling e.g. the displacement $u$ of a vibrating elastic string fixed at $x=0$ and $x=a$

## 2. What is Numerical Analysis?

Categories of mathematical problems

| Category | Algebra | Analysis |
| :--- | :---: | :---: |
| linear | computable | not computable |
| nonlinear | not computable | not computable |

Algebra Only finite constructs
Analysis Limits, derivatives, integrals etc. (transfinite)
Computable Exact solution obtained with finite computation

## What is Numerical Analysis?

Categories of mathematical problems

| Category | Algebra | Analysis |
| :--- | :---: | :---: |
| linear | computable | not computable |
| nonlinear | not computable | not computable |

Problems from all four categories can be solved numerically:

Numerical analysis aims to construct and analyze quantitative methods for the automatic computation of approximate solutions to mathematical problems

Goal Construction of mathematical software

## 3. The four principles of numerical analysis

There are four basic principles in numerical analysis. Every computational method is constructed from these principles

- Discretization

To replace a continuous problem by a discrete one

- Linear algebra tools, including polynomials All techniques that are computable
- Linearization

To approximate a nonlinear problem by a linear one

- Iteration

To apply a computation repeatedly (until convergence)

## Discretization

From function to vector




A continuous function is sampled at discrete points

## Discretization ...

## Reduces the amount of information to a finite set

Select a subset $\Omega_{N}=\left\{x_{0}, x_{1}, \ldots, x_{N}\right\}$ of distinct points from the interval $\Omega=[0,1]$

Continuous function $f(x)$ is replaced by vector $F=\left\{f\left(x_{k}\right)\right\}_{0}^{N}$

The subset $\Omega_{N}$ is called the grid

The vector $F$ is called a grid function or discrete function

Discretization yields a computable problem. Approximations are computed only at a finite set of grid points

## Discretization

## Approximation of derivatives

Definition of derivative

$$
f^{\prime}(x)=\lim _{\Delta x \rightarrow 0} \frac{f(x+\Delta x)-f(x)}{\Delta x}
$$

Take $\Delta x=x_{k+1}-x_{k}$ finite and approximate

$$
f^{\prime}\left(x_{k}\right) \approx \frac{f\left(x_{k}+\Delta x\right)-f\left(x_{k}\right)}{\Delta x}=\frac{F_{k+1}-F_{k}}{x_{k+1}-x_{k}}
$$

How accurate is this approximation?

## Error and accuracy

Expand in Taylor series

$$
\begin{aligned}
\frac{f\left(x_{k}+\Delta x\right)-f\left(x_{k}\right)}{\Delta x} & \approx \frac{f\left(x_{k}\right)+\Delta x f^{\prime}\left(x_{k}\right)+\mathcal{O}\left(\Delta x^{2}\right)-f\left(x_{k}\right)}{\Delta x} \\
& =f^{\prime}\left(x_{k}\right)+\mathcal{O}(\Delta x)
\end{aligned}
$$

The error is proportional to $\Delta x$

Try this formula for $f(x)=\mathrm{e}^{x}$ at $x_{k}=1$ for various $\Delta x$

## Numerical differentiation

Error in numerically calculated derivative of $\mathrm{e}^{x}$ at $x=1$


Loglog plot of error vs. step size $\Delta x$

Error in numerically calculated derivative of $\mathrm{e}^{x}$ at $x=1$


Loglog plot of error vs. step size $\Delta x$

## Better alternative

## Symmetric approximation

$$
\frac{f\left(x_{k}+\Delta x\right)-f\left(x_{k}-\Delta x\right)}{2 \Delta x}=\frac{F_{k+1}-F_{k-1}}{x_{k+1}-x_{k-1}}=f^{\prime}\left(x_{k}\right)+\mathcal{O}\left(\Delta x^{2}\right)
$$

The approximation is $2 n d$ order accurate because error is $\Delta x^{2}$

Whenever you have a power law, plot in log-log diagram

$$
\text { err } \approx C \cdot \Delta x^{p} \quad \Rightarrow \quad \log \mathrm{err} \approx \log C+p \log \Delta x
$$

This is a straight line of slope $p$, easy to identify

## Numerical 2nd order differentiation

Error in numerically calculated derivative of $\mathrm{e}^{x}$ at $x=1$


Loglog plot of error vs. step size $\Delta x$

## Second order approximation

## Good news, bad news

Error in numerically calculated derivative of $\mathrm{e}^{x}$ at $x=1$


Loglog plot of error vs. step size $\Delta x$

## 1st and 2nd order comparison

Error in numerically calculated derivative of $\mathrm{e}^{x}$ at $x=1$


2nd order approximation has much higher accuracy!

## Discretization of a differential equation

Problem Solve $\dot{y}=q y$ numerically on $[0, T]$, with $y(0)=y_{0}$

The analytical solution is a function $y(t)=y_{0} \mathrm{e}^{q t}$
The numerical solution is a vector $y=\left\{y_{k}\right\}_{k=0}^{N}$ with $y_{k} \approx y\left(t_{k}\right)$, where $t_{k}=k T / N$

Method Approximate derivative by difference quotient, using step size

$$
\Delta t=t_{k+1}-t_{k}=T / N
$$

## Discretization of a differential equation ...

Discretization method Approximate derivative by finite difference quotient

$$
\frac{y_{k+1}-y_{k}}{t_{k+1}-t_{k}} \approx \dot{y}\left(t_{k}\right)
$$

to get the linear algebraic equation system $(k=0,1, \ldots, N)$

$$
\frac{y_{k+1}-y_{k}}{t_{k+1}-t_{k}}=q y_{k}
$$

or

$$
y_{k+1}=\left(1+\frac{q T}{N}\right) y_{k} \quad \Rightarrow \quad y_{N}=\left(1+\frac{q T}{N}\right)^{N} y_{0}
$$

## Discretization of a differential equation...

Note The numerical approximation at time $T$ is

$$
y_{N}=\left(1+\frac{q T}{N}\right)^{N} y_{0}
$$

But

$$
\lim _{N \rightarrow \infty}\left(1+\frac{q T}{N}\right)^{N}=\mathrm{e}^{q T}
$$

The numerical approximation converges to the exact solution as $N \rightarrow \infty$

## Main course objective

The course centers on discretization methods for differential equations of all types

We will also encounter the other three principles of numerical analysis as we go

- linear algebra methods and polynomial techniques
- linearization
- and iterative methods

