

# Simulation and Analysis of 1D Wave Propagation under Various Physical Models

Dario Liotta



UNIVERSITÀ  
DEGLI STUDI  
DI PADOVA



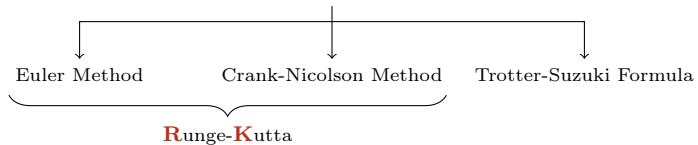
Dipartimento  
di Fisica  
e Astronomia  
Galileo Galilei

September 6th 2025

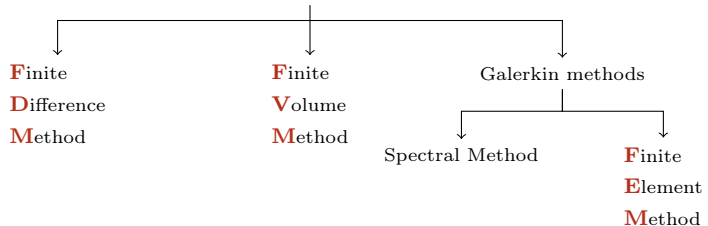
Course of **Quantum Information and Computing**  
Academic Year 2024/2025

# Numerical methods for differential equations

## Ordinary Differential Equations

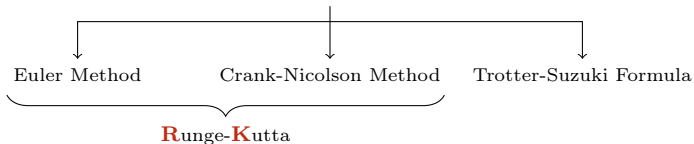


## Partial Differential Equations



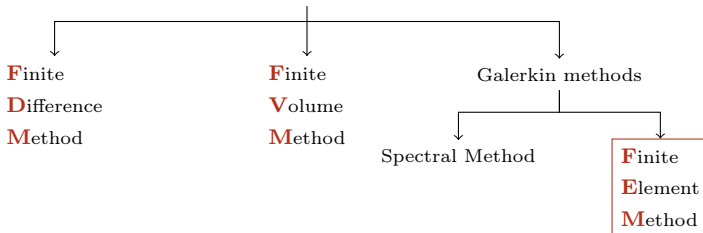
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## Partial Differential Equations



# Introduction to the problem

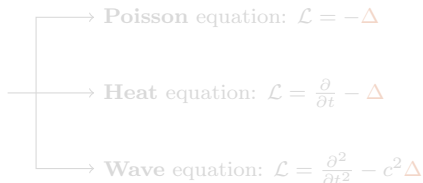
Solving a **PDE** means to find a function  $u$  such that

$$\mathcal{L}u = f$$

where  $\mathcal{L}$  is a differential operator and  $f$  is a source term.

The equation holds in a domain  $\Omega$  and is completed by prescribing **boundary conditions** on  $\partial\Omega$ .

In most physical  
applications  $\mathcal{L}$  is a  
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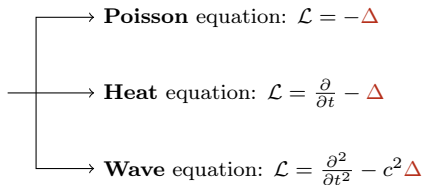
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# Weak formulation

Galerkin methods rely on a **weak formulation**

- Multiply by a **test function**  $v$  and integrate over the entire domain

$$-\int_{\Omega} (\Delta u) v d\Omega = \int_{\Omega} f v d\Omega$$

- Integrate by parts the left hand side

$$-\int_{\Omega} (\Delta u) v d\Omega = \int_{\Omega} \nabla u \cdot \nabla v d\Omega - \int_{\partial\Omega} \frac{\partial u}{\partial n} v ds$$

- Substitute and get the new expression

$$\int_{\Omega} \nabla u \cdot \nabla v d\Omega = \int_{\Omega} f v d\Omega + \int_{\partial\Omega} \frac{\partial u}{\partial n} v ds$$

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# About the test function

The test function  $v$  is introduced to check whether the PDE is satisfied on average throughout the domain.

The problem becomes to find  $u$  such that

$$a(u, v) = F(v) \quad \forall v \in V$$

where

$$a(u, v) = \int_{\Omega} \nabla u \cdot \nabla v d\Omega \quad \text{is a bilinear form}$$
$$F(v) = \int_{\Omega} f v d\Omega + \int_{\partial\Omega} \frac{\partial u}{\partial n} v ds \quad \text{is a linear functional}$$

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# Benefits of the weak formulation

## Strong formulation

$$u \in C^2(\Omega)$$

Holds pointwise in  $\Omega$

Derivatives exist classically

## Weak formulation

$$u, v \in H^1(\Omega)^*$$

Holds on average on  $\Omega$

Derivatives exist in the  
distributional sense

In short: weak formulation requires **less regularity**

\*  $H^1(\Omega)$  is a **Sobolev space** of functions with square-integrable first derivatives:

$$w \in H^1(\Omega) = \left\{ w \in L^2(\Omega) \mid \nabla w \in L^2(\Omega)^d \right\}$$

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Another difference lies in the boundary condition prescription.



$v = 0$  on  $\partial\Omega \Rightarrow$  cancels boundary term  
(no information available on  $\frac{\partial u}{\partial n}$ )

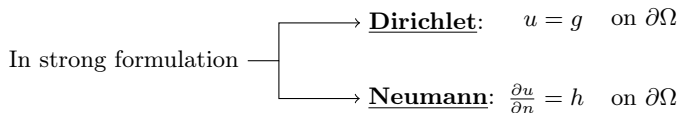
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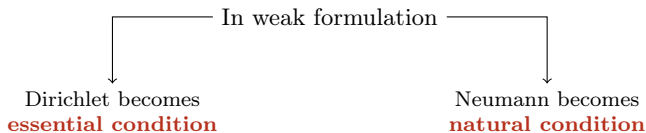
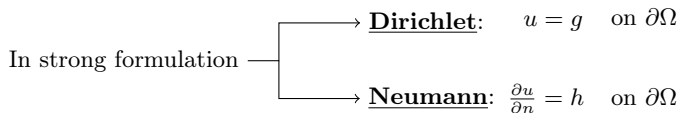
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# Shape functions

Galerkin methods allow to find an approximate solution

$$u_h \in V_h \subset H^1(\Omega) \quad \text{where } V_h \text{ is a **finite-dimensional** space}$$

In this framework, the goal is to find  $u_h$  such that

$$a(u_h, v_h) = F(v_h) \quad \forall v_h \in V_h$$

A **basis of function**  $\{\phi_i\}$  is chosen to express  $u_h$  and to use it as test:

$$u_h = \sum_{j=1}^N u_j \phi_j \implies a\left(\sum_{j=1}^N u_j \phi_j, \phi_i\right) = F(\phi_i) \quad \forall i = 1, \dots, N$$

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# Final expression

By linearity of  $a(\cdot, \cdot)$ , the problem reduces to a **finite linear system**:

$$\sum_{j=1}^N u_j a(\phi_j, \phi_i) = F(\phi_i) \quad \forall i = 1, \dots, N$$



$$\boxed{A\mathbf{u} = \mathbf{F}}$$

where

$$A_{i,j} = a(\phi_j, \phi_i)$$

$$\mathbf{u} = (u_1, \dots, u_N)^T$$

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is the **vector of unknowns**

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# Mesh discretization

**FEM** approach consists in the subdivision of the domain in a so-called **mesh**

This choice brings several advantages:

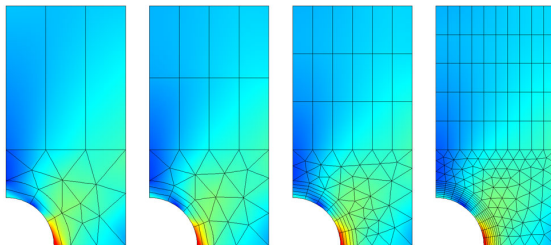
- Good approximation of **complex geometries**
- Better capture of **local effects**
- Possibility of **adaptive refinement**
- Natural construction of a **global solution**

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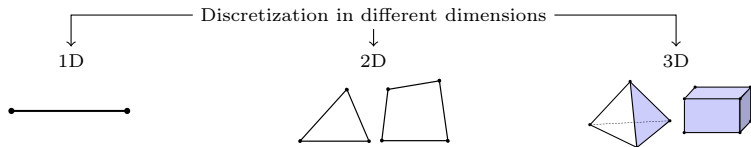
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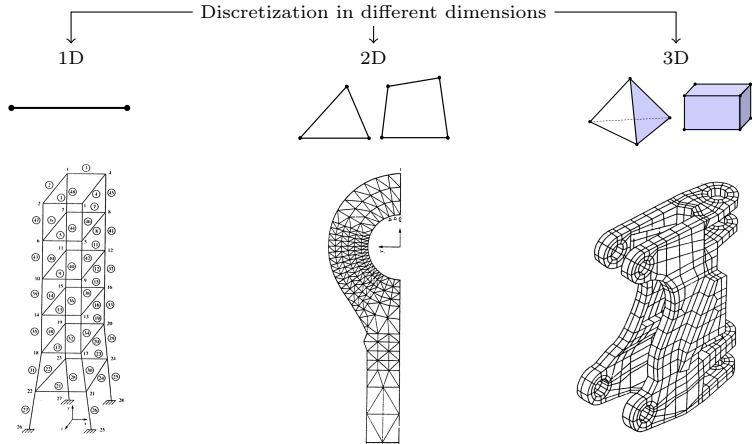


# Elements





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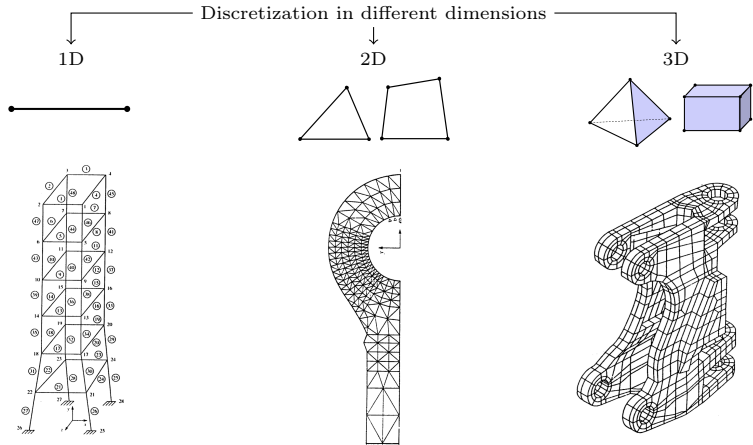


**Frame elements**

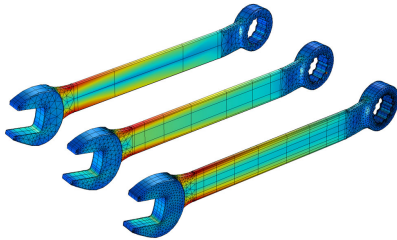
**Triangular elements**

**Brick elements**

# Elements

Frame **elements**Triangular **elements**Brick **elements**Finite **E**lement Method

# Application examples

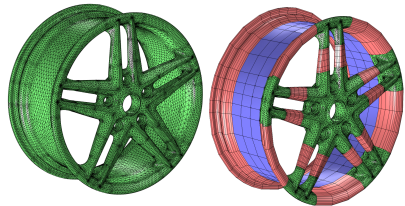


*Manual mesh refinement of a wrench using different element types*

Image from COMSOL Multiphysics Cyclopedica, "Finite Element Mesh Refinement", 21st of February 2017

*Mesh of a wheel rim composed of tetrahedrons in green, bricks in blue and prisms in pink*

Image from COMSOL Multiphysics Blog, "Meshing Your Geometry: When to Use the Various Element Types", Walter Frei, 4th of November 2013



# Choise of the base

Mesh division into sub-domains



Choice of a **local basis functions**



Exploitation of compact support



- Leads to **sparse matrices**
- Allows **local interpolation**
- Enhances **numerical stability**
- Enables **efficient parallelization**

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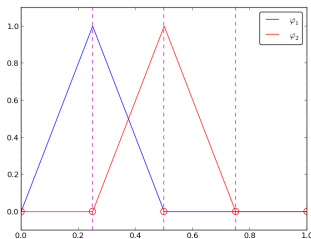
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# FEniCS library

A leading software platform for finite element computations is **FEniCS**.

- **Open-source** and freely available
- **Multi-language support** (C++ and Python APIs)
- **Parallel computing** with MPI support

FEniCS package: {

DOLFIN	(backend core engine and PETSc interface)
UFL	(symbolic language)
FIAT	(shape functions tabulator)
FFC	(C++ compiler for efficient local assembly)
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# A minimal FEniCS example: setup

Setup of a Poisson equation with Neumann boundary conditions in FEniCS:

- Generation of the mesh

```
domain = mesh.create_interval(MPI.COMM_WORLD, nx, [0.0, L])
```

- Definition of the finite element function space

```
V = functionspace(domain, ("Lagrange", 1))
```

- Definition of trial function and test function

```
u = ufl.TrialFunction(V)  
v = ufl.TestFunction(V)
```

- Definition of the source term

```
f = fem.Constant(domain, default_scalar_type(-6))
```

# A minimal FEniCS example: solution

Solving Poisson equation with Neumann boundary conditions in FEniCS:

- Weak formulation

```
a = ufl.dot(ufl.grad(u), ufl.grad(v)) * ufl.dx
F = f * v * ufl.dx
```

- Solution of the linear system

```
problem = LinearProblem(a, F,
                        petsc_option = {"ksp_type": "preonly",
                                         "pc_type" : "lu"
                                         }
                        )
u_h = problem.solve()
```

## Title

Our first goal is to approximate the solution of

$$\boxed{\frac{\partial^2 u}{\partial t^2} - c^2 \frac{\partial^2 u}{\partial x^2} = 0} \leftarrow \text{d'Alembert equation}$$

Equation is separable

Spatial part

$$-c^2 \int_0^L \frac{\partial^2 u}{\partial x^2} v dx =$$

Temporal part

$$\frac{\partial^2 u}{\partial t^2} \simeq \frac{u^{(n+1)} - 2u^{(n)} + u^{(n-1)}}{\Delta t^2}$$