





# Numerical Solution of PDEs Using the Finite Element Method May 15 – 19 2017

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## **Motivations**

# Many of the big problems in scientific computing are described by partial differential equations (PDEs):

- Structural statics and dynamics
  - Bridges, roads, cars, ...
- Fluid dynamics
  - Ships, pipe networks, ...
- Aerodynamics
  - Cars, airplanes, rockets, ...
- Plasma dynamics
  - Astrophysics, fusion energy
- But also in many other fields: Biology, finance, epidemiology, ...







## **Numerics for PDEs**

#### There are 3 standard tools for the numerical solution of PDEs:

- Finite element method (FEM)
- Finite volume method (FVM)
- Finite difference method (FDM)

#### **Common features:**

- Split the domain into small volumes (cells)
- Define balance relations on each cell
- Obtain and solve very large (non-)linear systems

#### **Problems:**

- Every code has to implement these steps
- There is only so much time in a day
- There is only so much expertise anyone can have







## **Ideal Characteristics**

### **Examples of what we would like to have:**

- Adaptive meshes
- Realistic, complex geometries
- Quadratic or even higher order elements
- Multigrid solvers
- Scalability to 1000s of processors
- Efficient use of current hardware
- Graphical output suitable for high quality rendering

Q: How can we make all of this happen in a single code?







## **Main Question!**

Q: How can we make all of this happen in a single code?

### Not a question of feasibility but of how we develop software:

- Is every student developing their own software?
- Or are we re-using what others have done?
- Do we insist on implementing everything from scratch?
- Or do we build our software on existing libraries?

There has been a major shift on how we approach the second question in scientific computing over the past 10-15 years!







## Main Answer...

The secret to good scientific software is (re)using existing libraries!







### There is excellent software for almost every purpose!

Basic linear algebra (dense vectors, matrices):

- BLAS
- LAPACK

Parallel linear algebra (vectors, sparse matrices, solvers):

- PETSc
- Trilinos

Meshes, finite elements, etc:

- deal.II the topic of this class
- ...

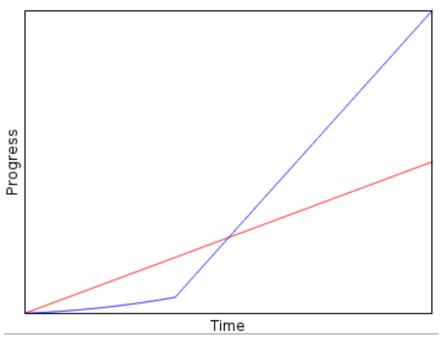
Visualization, dealing with parameter files, ...







## **Progress over time:**



Red: Do it yourself.

Blue: Use existing software.

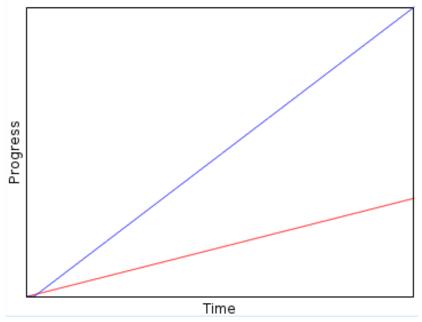
**Question:** Where is the cross-over point?







### **Progress over time, the real picture:**



Red: Do it yourself.

Blue: Use existing software.

**Answer:** Cross-over is after 2–4 weeks! A PhD takes 3–4 years.







### **Experience:**

It is realistic for a student developing numerical methods to have a code at the end of a PhD time that:

- Works in 2d and 3d
- On complex geometries
- Uses higher order finite element methods
- Uses multigrid solvers or preconditioners
- Solves a nonlinear, time dependent problem

Doing this from scratch would take 10+ years.







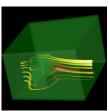
# Why Deal.II?

Publications per year

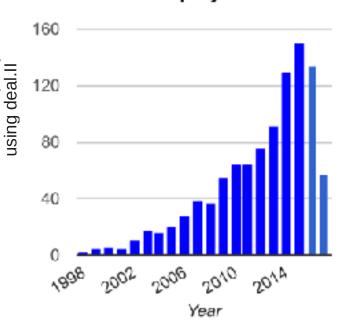
### deal.II is probably the largest FEM library:

- Presently ~600,000 lines of C++ code
- 10,000+ pages of documentation
- ~45 tutorial programs
- Fairly widely distributed:
   20,000+ downloads in 2012
- At least 65+ publications in 2012,
   996 overall, that use it
- Used in teaching at a number of universities
- 2007 Wilkinson prize.











# Why Deal.II?

### Linear algebra in deal.II:

- Has its own sub-library for dense + sparse linear algebra
- Interfaces to PETSC, Trilinos, UMFPACK

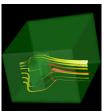


- Can read most mesh formats
- Can write almost any visualization file format

#### Parallelization:

- Uses threads and tasks on multicore machines
- Uses MPI, up to 10,000s of processors















# Why Deal.II?

### **Professional-level software management:**

- Globally accessible repository
- Mailing lists with significant volume
  - for user questions
  - for developer discussions
- ~2,700 tests run after every change
- Multi-platform build systems
  - Linux/Unix
  - Mac OS X
  - Windows
- Web sites tracking changes, tests, builds, ...







## **Prerequisites**

### **Operating system:**

- Linux or other Unix
- Mac OS X
- Windows

## C++ Compiler:

- GNU Compiler Collection (GCC)
- Intel C++
- Clang++

Other software: Perl, CMake, viz software, IDE

Disk space: 1-2GB







# Obtaining deal.II

### **Development sources**

Fork our repository on github:

https://github.com/dealii/dealii.git







# Building

### **Detailed instructions:**

http://www.dealii.org/developer/readme.html

### **Basic instructions:**

- Create build directory:
   cd deal.II; mkdir build; cd build
- Determine properties of the system and compilers: cmake -DCMAKE\_INSTALL\_PREFIX=/a/b/c ...
- Compile the entire library (~1 hour):
   make; make install
- Verify that everything works:

cd examples/step-1 cmake -DDEAL\_II\_DIR=/a/b/c . ; make run







## Building

There are many flags other than

- the installation directory
- the source directory

that can be given to *cmake*. For example:

- Paths to PETSc, Trilinos and other libraries
- Whether to build documentation locally
- Whether to disable or enable multithreading

### **Detailed instructions:**

http://www.dealii.org/developer/readme.html







# Brief re-hash of the FEM, using the Poisson equation:

We start with the strong form:

$$-\Delta u = f \quad \text{in } \Omega$$
  
$$u = 0 \quad \text{on } \partial \Omega$$







# Brief re-hash of the FEM, using the Poisson equation:

We start with the strong form:

$$-\Delta u = f$$

...and transform this into the weak form by multiplying from the left with a test function:

$$(\nabla \phi, \nabla u) = (\phi, f) \quad \forall \phi$$

The solution of this is a function u(x) from an infinite-dimensional function space.



Since computers can't handle objects with infinitely many coefficients, we seek a finite dimensional function of the form

$$u_h = \sum_{j=1}^N U_j \phi_j(x)$$

To determine the *N* coefficients, test with the *N* basis functions:

$$(\nabla \phi_i, \nabla u_h) = (\phi_i, f) \quad \forall i = 1...N$$

If basis functions are linearly independent, this yields *N* equations for *N* coefficients.

This is called the *Galerkin* method.



**Practical question 1:** How to define the basis functions?

**Answer:** In the finite element method, this is done using the following concepts:

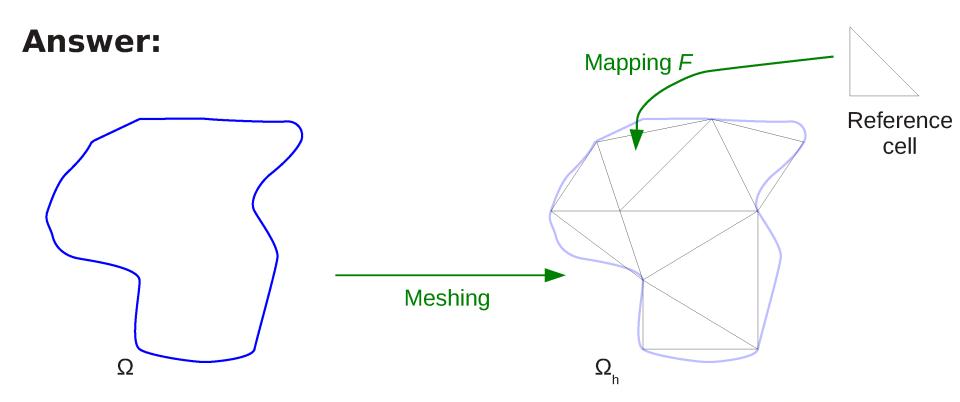
- Subdivision of the domain into a mesh
- Each cell of the mesh is a mapping of the reference cell
- Definition of basis functions on the reference cell
- Each shape function corresponds to a degree of freedom on the global mesh







**Practical question 1:** How to define the basis functions?



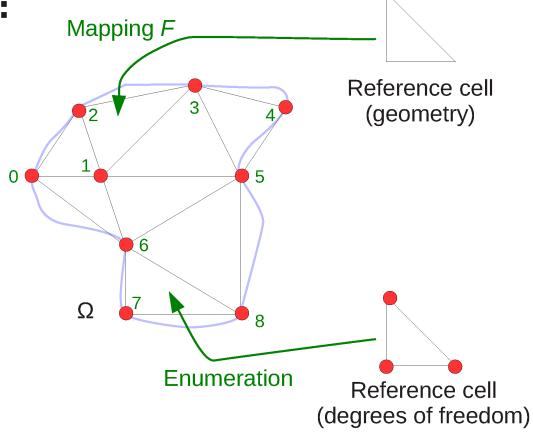






## **Practical question 1:** How to define the basis functions?

### **Answer:**





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- Subdivision of the domain into a mesh
- Each cell of the mesh is a mapping of the reference cell
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- Each shape function corresponds to a degree of freedom on the global mesh

Concepts in red will correspond to things we need to implement in software, explicitly or implicitly.



Given the definition  $u_h = \sum_{j=1}^N U_j \phi_j(x)$ , we can expand the bilinear form

$$(\nabla \phi_i, \nabla u_h) = (\phi_i, f) \quad \forall i = 1...N$$

to obtain:

$$\sum_{i=1}^{N} (\nabla \phi_i, \nabla \phi_j) U_j = (\phi_i, f) \quad \forall i = 1...N$$

This is a linear system

$$AU=F$$

with

$$A_{ij} = (\nabla \phi_i, \nabla \phi_j)$$
  $F_i = (\phi_i, f)$ 







### Practical question 2: How to compute

$$A_{ij} = (\nabla \phi_i, \nabla \phi_j) \qquad F_i = (\phi_i, f)$$

**Answer:** By mapping back to the reference cell...

$$A_{ij} = (\nabla \phi_i, \nabla \phi_j)$$

$$= \sum_{K} \int_{K} \nabla \phi_i(x) \cdot \nabla \phi_j(x)$$

$$= \sum_{K} \int_{\hat{K}} J_K^{-1}(\hat{x}) \hat{\nabla} \hat{\phi}_i(\hat{x}) \cdot J_K^{-1}(\hat{x}) \hat{\nabla} \hat{\phi}_j(\hat{x}) |\det J_K(\hat{x})|$$

...and quadrature:

$$A_{ij} \approx \sum_{K} \sum_{q=1}^{Q} J_{K}^{-1}(\hat{x}_{q}) \hat{\nabla} \hat{\varphi}_{i}(\hat{x}_{q}) \cdot J_{K}^{-1}(\hat{x}_{q}) \hat{\nabla} \hat{\varphi}_{j}(\hat{x}_{q}) \underbrace{|\det J(\hat{x}_{q})| \ w_{q}}_{=: \text{IxW}}$$

Similarly for the right hand side *F*.







**Practical question 3:** How to store the matrix and vectors of the linear system

$$AU = F$$

### **Answers:**

- A is sparse, so store it in compressed row format
- U,F are just vectors, store them as arrays
- Implement efficient algorithms on them, e.g. matrixvector products, preconditioners, etc.
- For large-scale computations, data structures and algorithms must be parallel







Practical question 4: How to solve the linear system

$$AU = F$$

Answers: In practical computations, we need a variety of

- Direct solvers
- Iterative solvers
- Parallel solvers







**Practical question 5:** What to do with the solution of the linear system

$$AU = F$$

**Answers:** The goal is not to solve the linear system, but to do something with its solution:

- Visualize
- Evaluate for quantities of interest
- Estimate the error

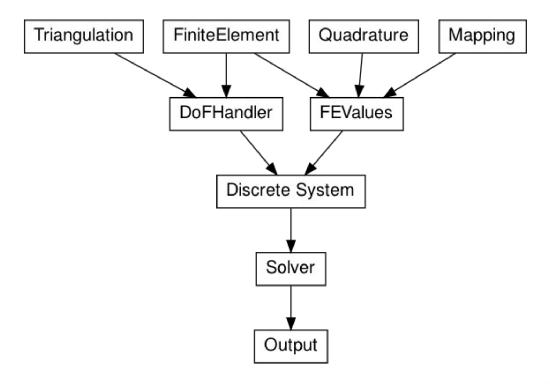
These steps are often called *postprocessing the solution*.







Together, the concepts we have identified lead to the following components that all appear (explicitly or implicitly) in finite element codes:









## After reading, play with the program:

cd examples/step-1 cmake -DDEAL\_II\_DIR=/path/to/deal.II . make run

This will run the program and generate output files:

ls -l okular grid-2.eps

**Next step:** Play by following the suggestions in the results section. This is the best way to learn!







## **Step-2 shows:**

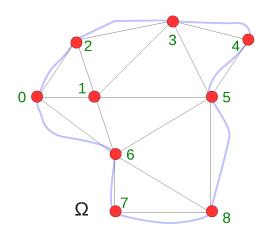
- How degrees of freedom are defined with finite elements
- The DoFHandler class
- How DoFs are connected by bilinear forms
- Sparsity patterns of matrices
- How to visualize a sparsity pattern







### **Example:** Consider this mesh and bilinear form:



$$A_{ij} = (\nabla \phi_i, \nabla \phi_j)$$
  
= 
$$\int_{\Omega} \nabla \phi_i \cdot \nabla \phi_j \, dx$$

**Note:** In general we have that

- $A_{00} \neq 0, A_{01} \neq 0, A_{02} \neq 0, A_{06} \neq 0$
- $A_{03} = A_{04} = A_{05} = A_{07} = A_{08} = 0$

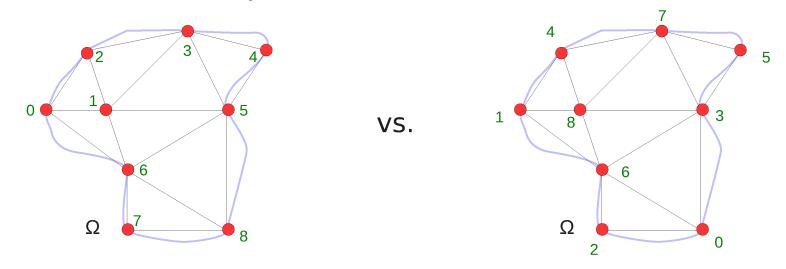
The bigger the mesh, the more zeros there are per row!







**Renumbering:** The order of enumerating degrees of freedom is arbitrary



### **Notes:**

- Resulting matrices are just permutations of each other
- Both sparse, but some algorithms care







## Step-3 shows:

- How to set up a linear system
- How to assemble the linear system from the bilinear form:
  - The loop over all cells
  - The *FEValues* class
- Solving linear systems
- Visualizing the solution







### **Recall:**

 For the Laplace equation, the bilinear form is written as a sum over all cells:

$$A_{ij} = (\nabla \phi_i, \nabla \phi_j)$$
  
=  $\sum_{K} \int_{K} \nabla \phi_i(x) \cdot \nabla \phi_j(x)$ 

- But on each cell, only few shape functions are nonzero!
- For  $Q_1$ , only  $16=4^2$  matrix entries are nonzero per cell
- Only compute this (dense) sub-matrix, then "distribute" it to the global A
- Similar for the right hand side vector.







### **Recall:**

We use quadrature

$$A_{ij}^{K} = \int_{K} \nabla \hat{\phi}_{i}(x) \cdot \nabla \hat{\phi}_{j} dx$$

$$\approx \sum_{q=1}^{Q} J_{K}^{-1}(\hat{x}_{q}) \hat{\nabla} \hat{\phi}_{i}(\hat{x}_{q}) \cdot J_{K}^{-1}(\hat{x}_{q}) \hat{\nabla} \hat{\phi}_{j}(\hat{x}_{q}) \underbrace{|\det J(\hat{x}_{q})| \ w_{q}}_{=:JxW}$$

- We really only have to evaluate shape functions,
   Jacobians, etc., at quadrature points not as functions
- All evaluations happen on the reference cells







# Now ... Exercise Time!







