

# Computational mathematics for heterogeneous materials

Axel Målqvist

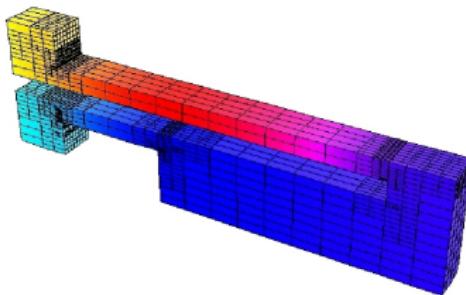
Department of Mathematical Sciences  
Chalmers University of Technology and University of Gothenburg

2018-05-17

# Partial differential equations (PDE)

Heat equation:

$$\frac{\partial}{\partial t} u(x, t) - \sum_{i=1}^3 \frac{\partial}{\partial x_i} \left( A(x) \frac{\partial u(x, t)}{\partial x_i} \right) = f(x, t)$$



Many physical processes are described mathematically by PDEs.

- Elasticity (structural mechanics, modal analysis)
- Electromagnetism (waves, Joule heating)
- Fluid flow (porous media, fluid-solid interaction)
- Quantum physics (wave function)

# Why solve partial differential equations?

Industry:

- Simulation is cheaper than experiments
- Systematic way to optimize design

Government:

- Environment (climate, radioactive waste)
- Disasters (flood waves, hurricanes, nuclear disaster)

Science:

- Detailed understanding of physical processes
- Build intuition in mathematical analysis (conjecture)

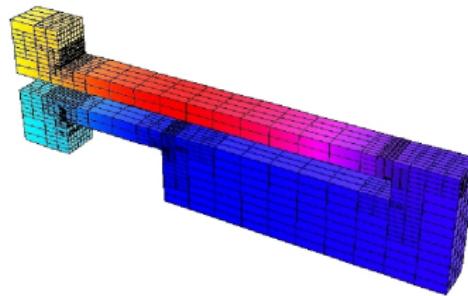
# Mathematical analysis and numerical analysis

Mathematical analysis can give:

- Existence, uniqueness
- Properties such as conservation of mass or energy
- Regularity (smoothness) of the solution
- Restrictions on the data

Numerical analysis uses all this to:

- Develop efficient algorithms to solve the PDE
- Analyze the error due to finite precision

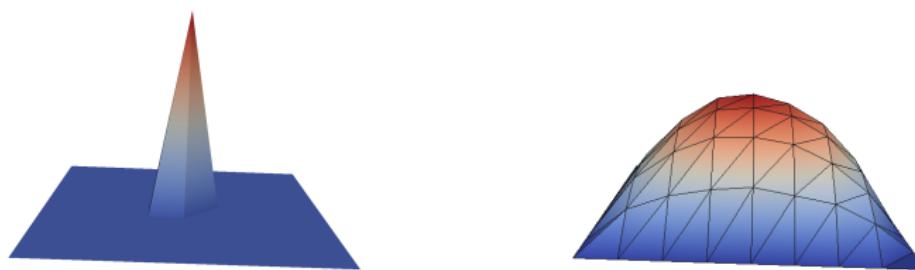


# Finite element method

Stationary heat equation:

$$-\sum_{i=1}^3 \frac{\partial}{\partial x_i} \left( A(x) \frac{\partial u(x)}{\partial x_i} \right) = f(x), \quad \text{in } \Omega.$$

with boundary condition  $u_{\partial\Omega} = 0$ . Let  $V_h \subset V$  be space of continuous piecewise linear functions.



The finite element solution  $u_h$  is an approximation (projection) of the exact solution.

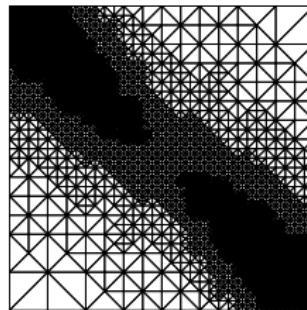
# Accuracy and challenges

The error for smooth data fulfills,

$$\|u - u_h\| := \left( \int_{\Omega} |u - u_h|^2 dx \right)^{1/2} \leq Ch^2 \|f\|,$$

here  $h$  is mesh size. This does *not* hold if

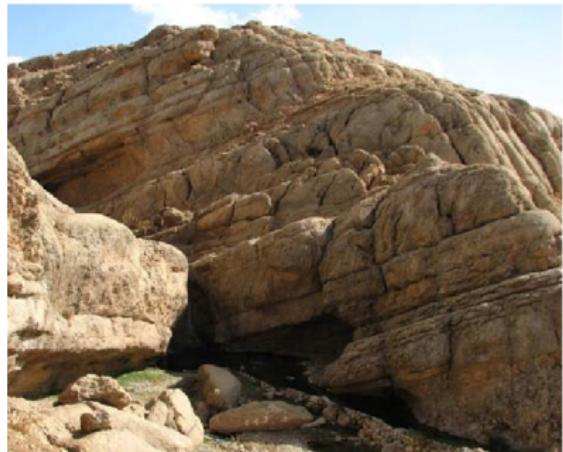
- Local roughness (corners, localized data  $f$ ),



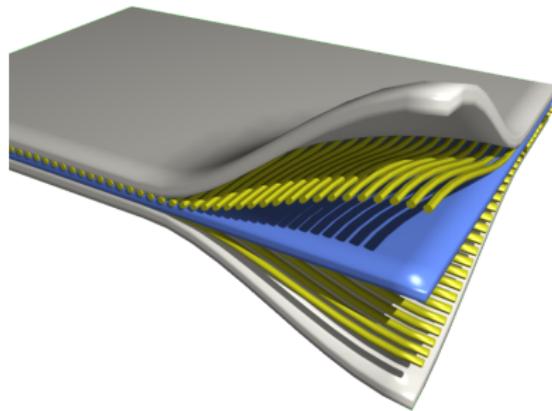
Local roughness is treated by adaptive mesh refinement.

- Global roughness (rough material data  $A$ ).

# Heterogeneous materials



Porous media flow



Composites

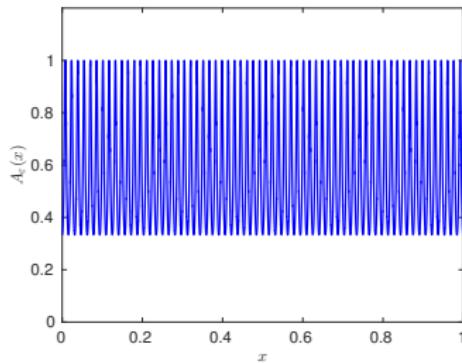
# Simple 1D example of how FEM fails

Example (periodic coefficient):

We consider

$$-\frac{d}{dx} \left( A(x) \frac{d}{dx} u(x) \right) = 1, \quad u(0) = u(1) = 0,$$

with  $A(x) = (2 + \cos(2\pi x/\varepsilon))^{-1}$ , where  $\varepsilon = 2^{-6}$ .



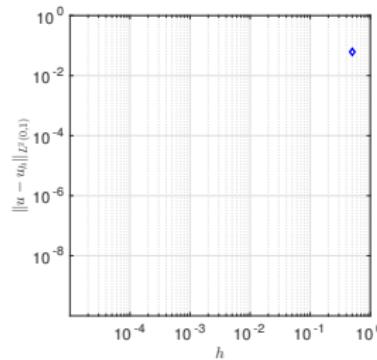
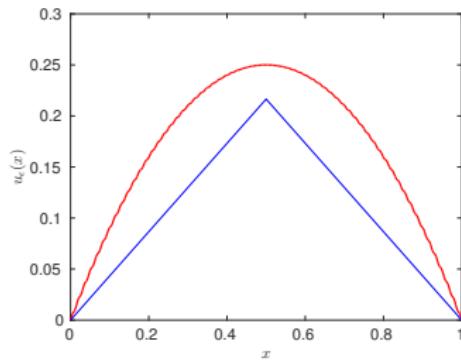
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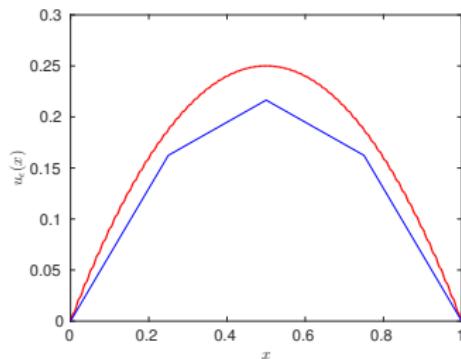
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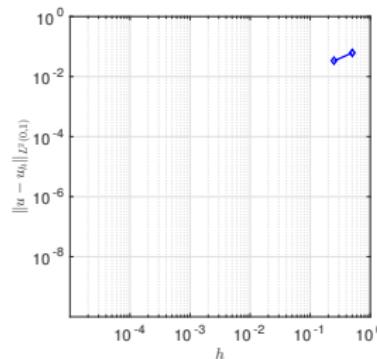
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solution and P1-FEM-approximation



$L^2(\Omega)$  – error vs.  $h$

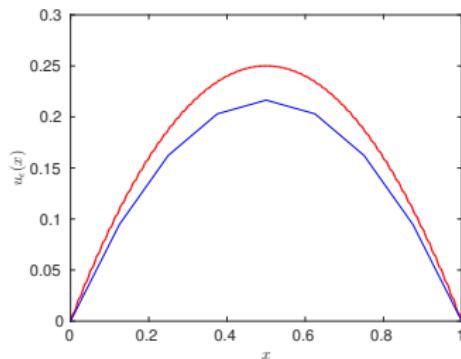
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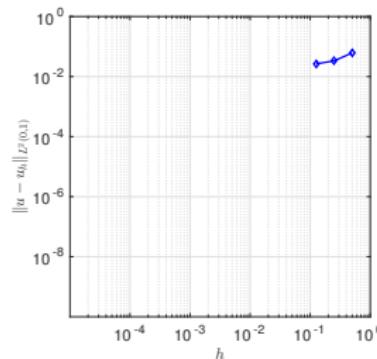
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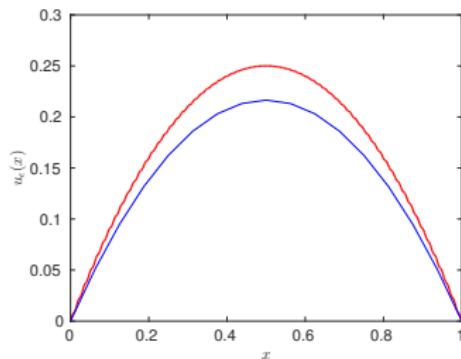
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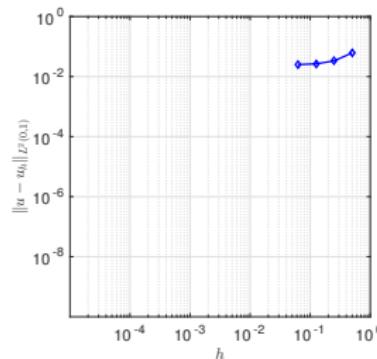
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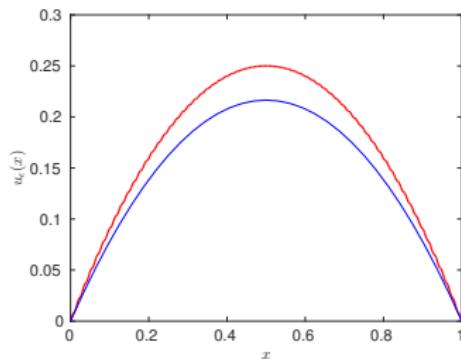
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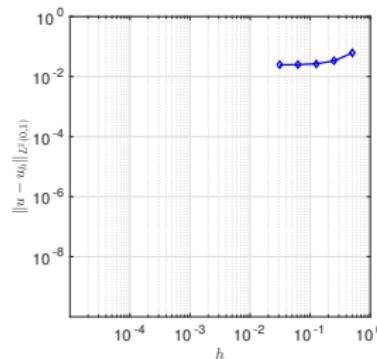
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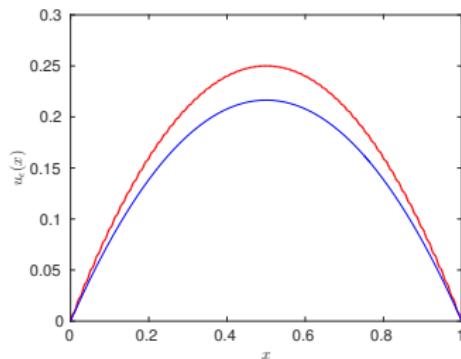
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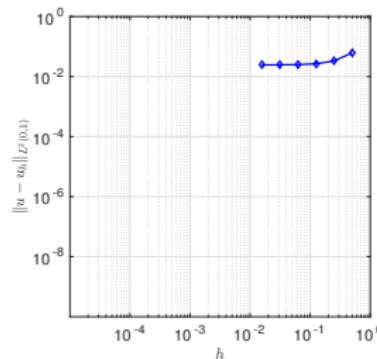
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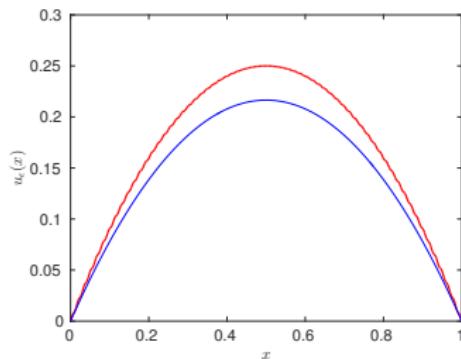
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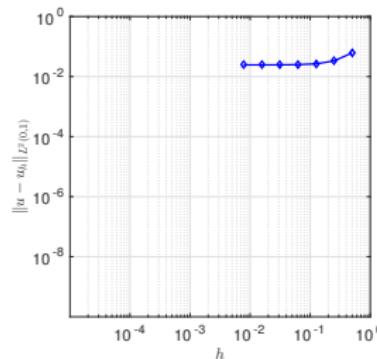
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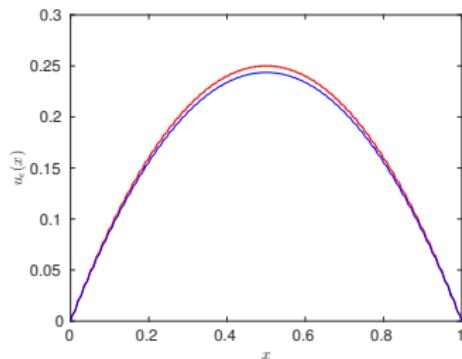
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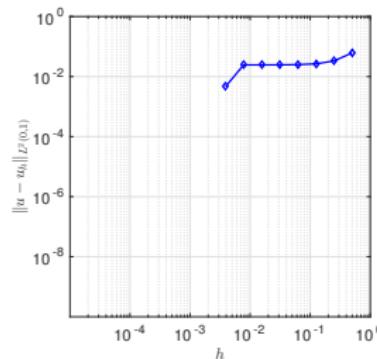
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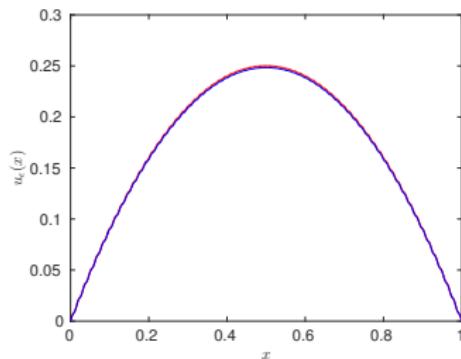
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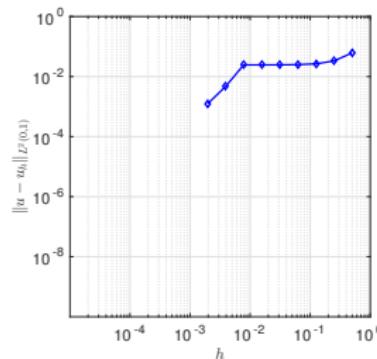
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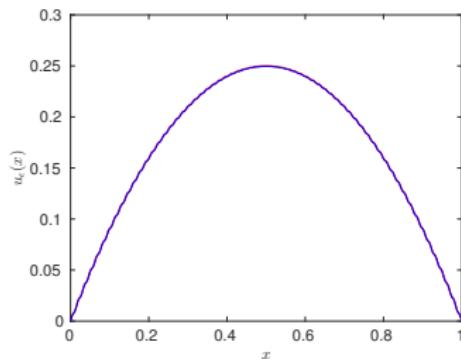
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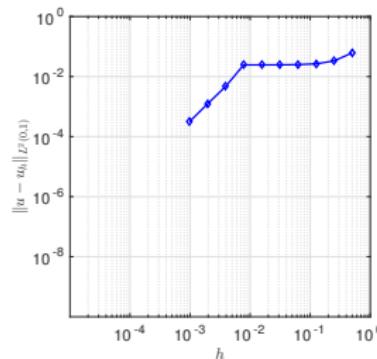
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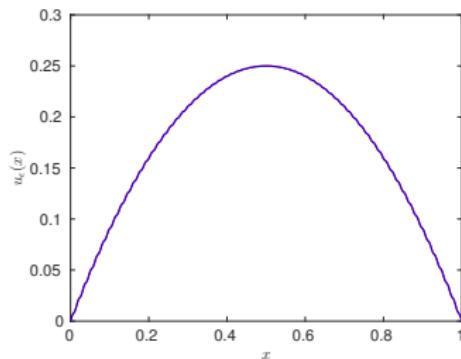
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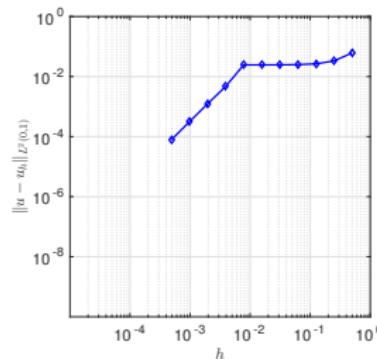
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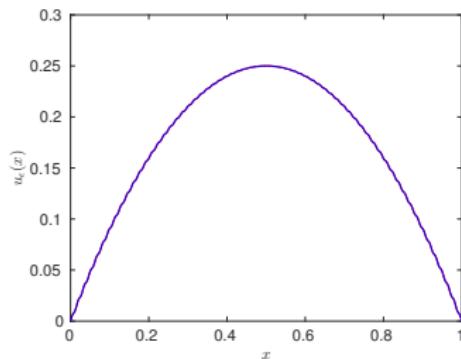
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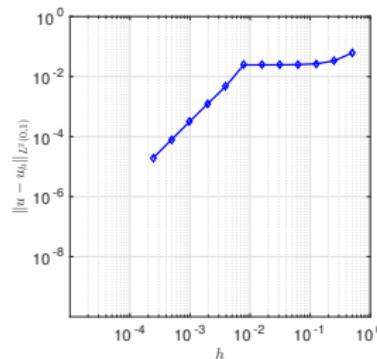
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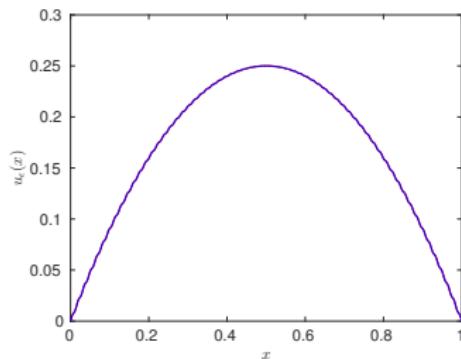
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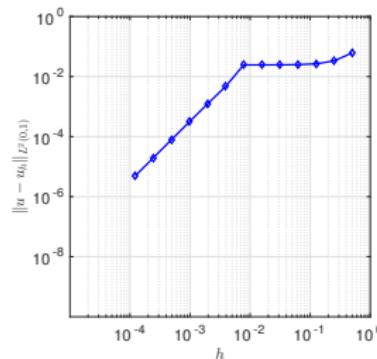
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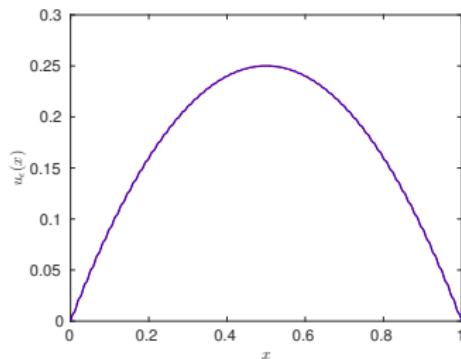
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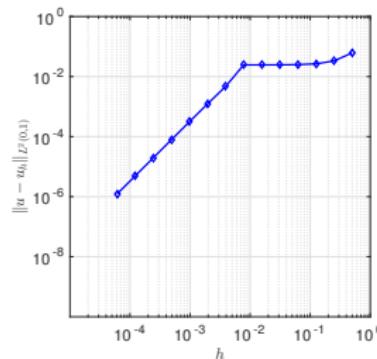
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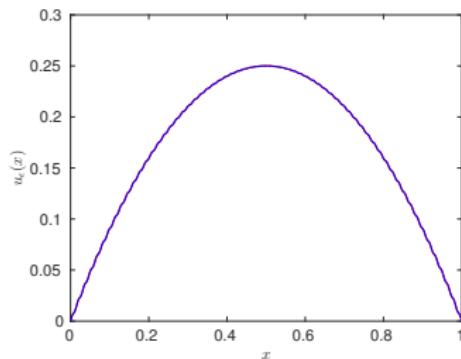
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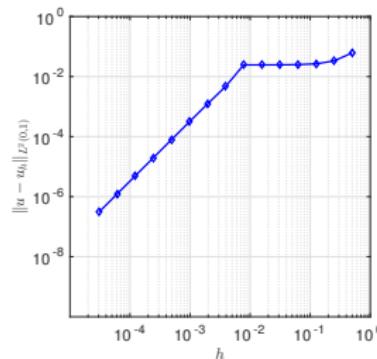
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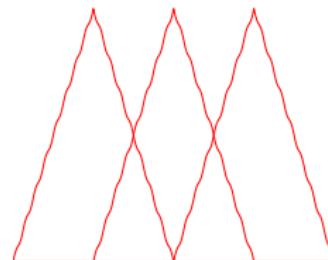
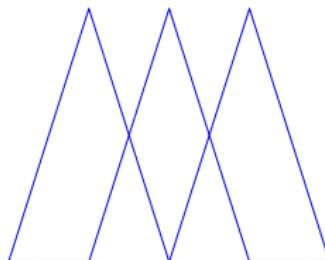


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# Resolution using two scale basis functions

In 1D one can construct two scale basis functions

$$\tilde{\Lambda}_j(x) := \begin{cases} \frac{\int_{x_{j-1}}^x A^{-1}(s) ds}{\int_{x_{j-1}}^{x_j} A^{-1}(s) ds}, & \text{if } x \in [x_{j-1}, x_j], \\ 1 - \frac{\int_{x_j}^x A^{-1}(s) ds}{\int_{x_j}^{x_{j+1}} A^{-1}(s) ds}, & \text{if } x \in [x_j, x_{j+1}], \\ 0, & \text{else.} \end{cases}$$

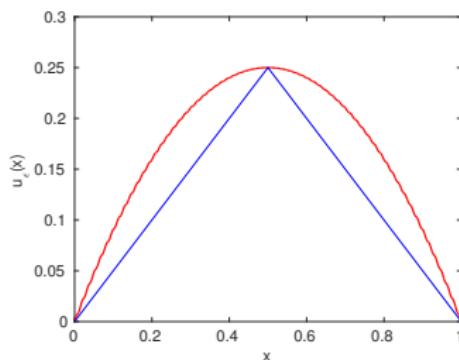


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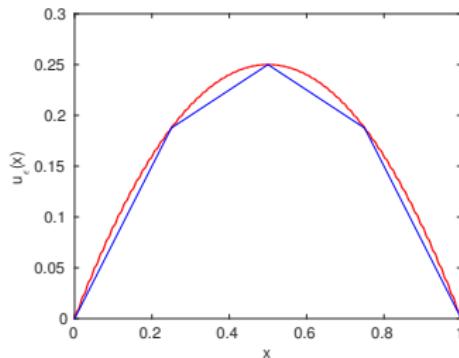


We plot  $I_h \tilde{u}_h$  which in fact is exact in the nodes.

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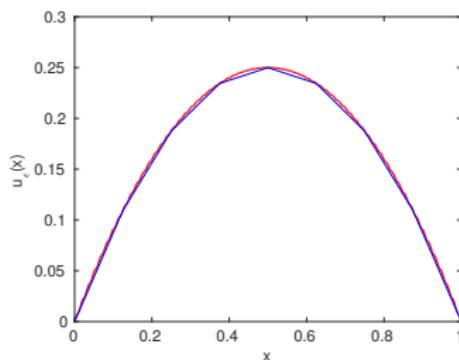


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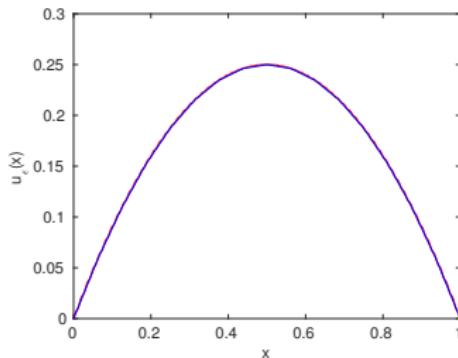


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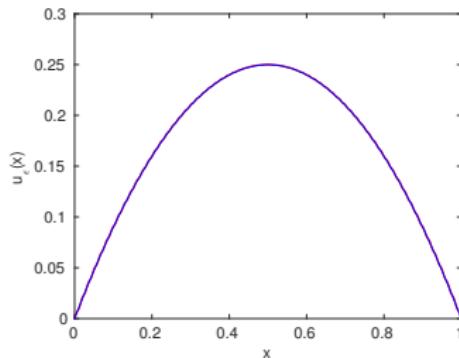


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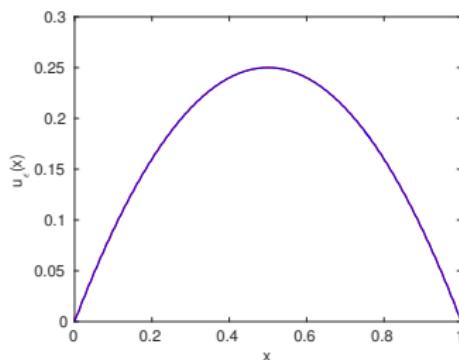


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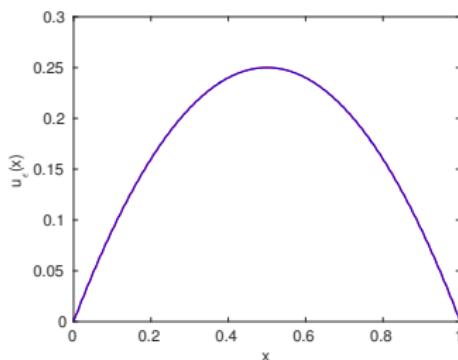


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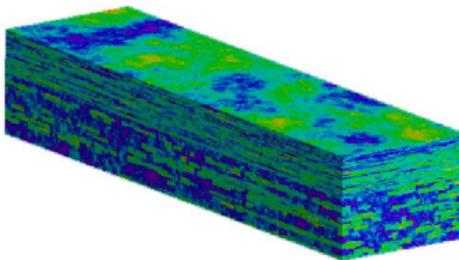
In 1D one can construct two scale basis functions

$$\tilde{\Lambda}_j(x) := \begin{cases} \frac{\int_{x_{j-1}}^x A^{-1}(s) ds}{\int_{x_{j-1}}^{x_j} A^{-1}(s) ds}, & \text{if } x \in [x_{j-1}, x_j], \\ 1 - \frac{\int_{x_j}^x A^{-1}(s) ds}{\int_{x_j}^{x_{j+1}} A^{-1}(s) ds}, & \text{if } x \in [x_j, x_{j+1}], \\ 0, & \text{else.} \end{cases}$$



We plot  $I_h \tilde{u}_h$  which in fact is exact in the nodes.

# Generalization to higher dimensions



- Resolving all scales with FEM is often too expensive
- No simple formula for  $\tilde{\Lambda}_j$  available
- Classical results only for periodic coefficients (homogenization theory)

My goal has been to construct optimal basis  $\tilde{\Lambda}_j$  without assuming periodicity (in 2D, 3D).

# Generalization to higher dimensions

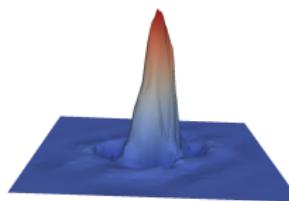
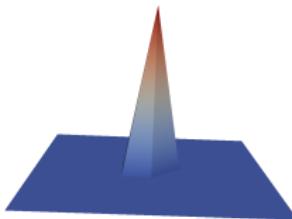
Poisson's equation on weak form: find  $u \in V$  such that

$$a(u, v) := (\mathbf{A}\nabla u, \nabla v) = (f, v), \quad v \in V.$$

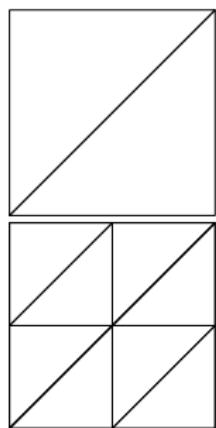
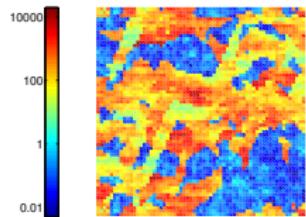
Let  $V = \tilde{V}_h \oplus V^f$  where  $V^f = \{v \in V : I_h v = 0\}$  and  $\tilde{V}_h = \{v \in V : a(v, w) = 0, \quad \forall w \in V^f\}$ . Let  $\tilde{u}_h \in \tilde{V}_h$  solve

$$a(\tilde{u}_h, v) = (f, v) = a(u, v), \quad v \in \tilde{V}_h.$$

Then  $I_h(u - \tilde{u}_h) = 0$  and furthermore  $\|u - \tilde{u}_h\| \leq Ch^2\|f\|$ .

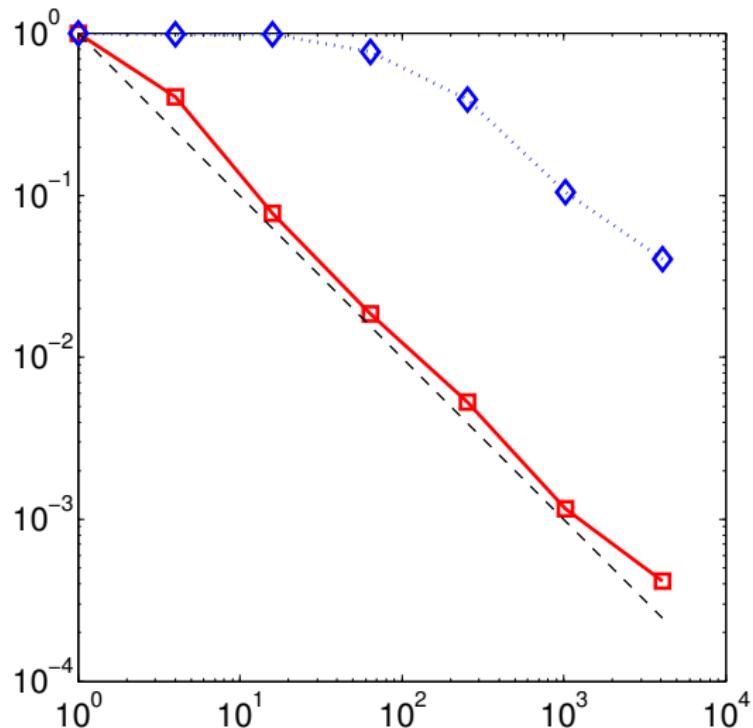


# Numerical experiment: Poisson's equation



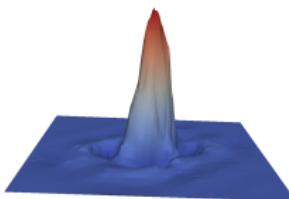
$$h = 2^{-1}, 2^{-2}, \dots, 2^{-7}$$

$$h_{\text{ref}} = 2^{-9}, \ell = \log(1/h)$$



$\|u_{h_{\text{ref}}} - \tilde{u}_h\|$  vs. #dof

# Mathematical challenges



- Prove exponential decay of basis
- High contrast data, channels, interpolation
- Prove a priori error bound independent of variations in  $A$
- Extend theory to problems relevant in applications (time dependent, nonlinear, systems)

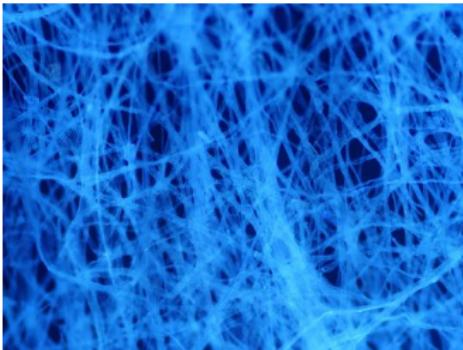
A. Persson, Numerical Analysis of Evolution Problems in Multiphysics, 2018.

# Recent developments

Several groups continues to develop this methodology  
(2014–present)

- Göteborg (time dependent, network models, high contrast data)
- Augsburg (Helmholtz, stochastic, quantum physics)
- Stockholm (Elasticity, quantum physics)
- Münster (Helmholtz, RBM)
- Berlin (iterative solvers)
- Lusanne (Wave)
- Heidelberg (High contrast data)
- Laramie (WY) (porous media flow)
- Athens (Discontinuous Galerkin)

# Future work



- Mechanical properties of paper (Fraunhofer)
- Random defects in materials
- Interface models for cracks, reinforcements