

Scientific Computing

Lecture 5

Differential equations (part 2)

Partial Differential Equations (PDE)

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Structure of the lecture

- ▶ Galerkin approach
- ▶ Elliptic equations
- ▶ Finite Volumes Method
- ▶ Parabolic equations
- ▶ Hyperbolic equations
- ▶ **EXAMPLE:** 3D Cauchy problem for elliptic equation

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Part 1: Galerkin approach

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The operator equation for differential equations is often written as:

$$D\mathbf{u} = 0,$$

where D represents both mathematical and the right-hand side (observed data); the function \mathbf{u} is an unknown function to be found.

- ▶ **Ordinary differential equation (ODE):** Let $x \in \mathbb{R}^1$. ODE of the k -th order can be represented with the operator:

$$D\mathbf{u} = F\left(x, u(x), u'(x), u''(x), \dots, u^{(k)}(x)\right).$$

- ▶ **Partial differential equation (PDE):** Let $\mathbf{x} \in \mathbb{R}^n \equiv (x_1, x_2, \dots, x_n)$. PDE of the k -th order can be represented with the following operator:

$$D\mathbf{u} = F\left(\mathbf{x}, \mathbf{u}(\mathbf{x}), \frac{\partial \mathbf{u}}{\partial x_1}, \dots, \frac{\partial \mathbf{u}}{\partial x_n}, \dots, \frac{\partial^2 \mathbf{u}}{\partial x_1^2}, \dots, \frac{\partial^2 \mathbf{u}}{\partial x_n^2}, \dots, \frac{\partial^{(k)} \mathbf{u}}{\partial x_1^{(k)}}, \frac{\partial^{(k)} \mathbf{u}}{\partial x_n^{(k)}}\right).$$

The Galerkin Method

- ▶ The Galerkin method may be considered an initial point - an **IDEA** - for linear DE/IE solution
- ▶ The Method is a base for wavelets analysis, FEM, FDM etc
- ▶ The Method shows one of the most general approaches in computational problems
- ▶ The Method allows to chose the concrete solution spaces with respect to the problem (may be considered a regularization too)
- ▶ Easy to understand
- ▶ Easy to research

The Galerkin Method

Consider the Boundary Value problem for the linear differential operator D :

$$Du = 0, \quad \text{in } \Omega \subset \mathbb{R}^n, \quad S(u) = 0, \text{ in } \partial\Omega$$

- ▶ Assume $u \in U$, $D : U \rightarrow V$, where U, V are some Hilbert spaces.
- ▶ Let $\{\varphi_i(\mathbf{x})\}, i = 1, \dots$ be some basis in U .
- ▶ Choose finite number N and approximate

$$u(\mathbf{x}) \approx u_a(\mathbf{x}) = u_0(\mathbf{x}) + \sum_{i=1}^N a_i \varphi_i(\mathbf{x}).$$

The Galerkin Method

Consider the Boundary Value problem for the linear differential operator D :

$$Du = 0, \quad \text{in } \Omega \subset \mathbb{R}^n, \quad S(u) = 0, \text{ in } \partial\Omega$$

- ▶ Assume $u \in U$, $D : U \rightarrow V$, where U, V are some Hilbert spaces.
- ▶ Let $\{\varphi_i(\mathbf{x})\}, i = 1, \dots$ be some basis in U .
- ▶ Choose finite number N and approximate the solution u
- ▶ Since N is finite, and u_a is only the approximation, we get:

$$D\left(\sum_{i=1}^N a_i \varphi_i(\mathbf{x})\right) + Du_0 = R(a_1, \dots, a_n, \mathbf{x}).$$

The Galerkin Method

Consider the Boundary Value problem for the linear differential operator D :

$$Du = 0, \quad \text{in } \Omega \subset \mathbb{R}^n, \quad S(u) = 0, \text{ in } \partial\Omega$$

- ▶ Assume $u \in U$, $D : U \rightarrow V$, where U, V are some Hilbert spaces.
- ▶ Let $\{\varphi_i(\mathbf{x})\}, i = 1, \dots$ be some basis in U .
- ▶ Choose finite number N and approximate the solution u
- ▶ Since N is finite, and u_a is only the approximation, we get the residual $R(a_1, \dots, a_n, \mathbf{x})$
- ▶ The function R is never zero... But we can minimize it with respect to the coefficients $\{a_i\}$!

- ▶ Represent (or approximate) the solution u with some weighted sum of functions
- ▶ Substitute the approximation or representation into the original equations
- ▶ Minimize the residual with respect to weights
- ▶ After the weights are calculated, reconstruct the approximate (or, sometimes, exact) solution, substituting the weights into your representation (or approximation) of it
- ▶ Be careful: the residual rarely being equal to zero, but it should be small!

Let the problem be linear: $D, S(u)$ are linear.

$$Du = 0, \quad \text{in } \Omega \subset \mathbb{R}^n, \quad S(u) = 0, \text{ in } \partial\Omega$$

- ▶ The weak formulation:

$$(Du, v) = 0, \quad v \in U.$$

- ▶ Approximate the solution:

$$u = \sum_{i=0}^N u_i \varphi_i(\mathbf{x}), \quad \varphi_i(\mathbf{x}) = \begin{cases} 1, & \mathbf{x} = \mathbf{x}_i \\ 0, & \mathbf{x} = \mathbf{x}_j, j \neq i \\ \text{continuous} & \text{elsewhere} \end{cases}$$

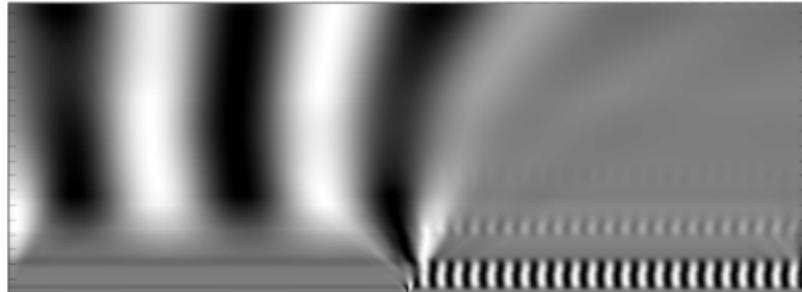
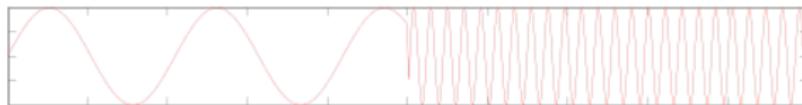
- ▶ Substitute the approximation into the weak form:

$$\sum_{i=1}^N u_i (D\varphi_i, \varphi_j) = 0, \quad \sum_{I: \mathbf{x}_I \in \partial\Omega} u_I S(\varphi_I(\mathbf{x})) = 0.$$

Continuous Wavelet Transform

With the wavelets we can analyse not only frequencies (like in Fourier analysis), but both frequencies and its locations in time (scaling and shifting)

$$U(a, b) = \frac{1}{|a|^{\frac{1}{2}}} \int_{\mathbb{R}} u(x) \psi^*\left(\frac{t-b}{a}\right) dt$$



- ▶ Continuous wavelet transform:

$$U(a, b) = \frac{1}{|a|^{\frac{1}{2}}} \int_{\mathbb{R}} u(x) \psi^*\left(\frac{t-b}{a}\right) dt$$

- ▶ The continuous wavelet transform (the inverse one, 1D case):

$$u(x) = C_{psi} \int_{\mathbb{R}^2} \frac{1}{a^2} U(a, b) \tilde{\psi}\left(i \frac{t-b}{a}\right) db da$$

- ▶ In discrete case:

$$u(x) = C_{\psi} \sum_{i,j=-\infty}^{\infty} U_{ij} \psi_{ij}(t), \quad U_{ij} = \int_{\mathbb{R}} u(x) \psi_{ij}^* dt$$

- ▶ In n -dimensional case we use separable wavelets: each for one dimension.

Multiresolution analysis

Let $\{V_j\}_{j=-\infty}^{\infty}$ is a sequence of spaces such that:

- ▶ $V_j \subset V_{j+1}$
- ▶ $\bigcup_{j \in \mathbb{Z}} V_j = L^2(\mathbb{R}), \quad \bigcap_{j \in \mathbb{Z}} V_j = \emptyset$
- ▶ $f(t) \in V_j \Rightarrow f(2t) \in V_{j+1}, \quad f(t) \in V_j \Rightarrow f(t - k) \in V_j$
- ▶ The mother wavelet ψ defines orthonormal basis in V_j :

$$\psi_{jk} = 2^{j/2} \psi(2^j t - k)$$

- ▶ Example: Haar's multiresolution analysis:

$$V_j = \{f \in L^2(\mathbb{R}); \forall k \in \mathbb{Z} : f|_{[2^j k, 2^j(k+1)]} = \text{const}\}$$

- ▶ After we choose the basis, then do the same: substitute the solution approximated with that basis!

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Part 2: Elliptic equations & FVM

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Partial differential equations of the second order

Partial differential equations of second order

Most of PDE-driven problems can be described with PDEs of second order. The differential operator D can be represented as follows:

$$D\mathbf{u} = \sum_{i=1}^n \sum_{j=1}^n a_{ij}(\mathbf{x}) \frac{\partial^2 \mathbf{u}}{\partial x_i \partial x_j} + F(\mathbf{x}, \mathbf{u}, \frac{\partial \mathbf{u}}{\partial x_1}, \dots, \frac{\partial \mathbf{u}}{\partial x_n}) \equiv L\mathbf{u} + F(\dots).$$

The operator L is called the Principle Part and being defined with the matrix:

$$L\mathbf{u} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ a_{n1} & \dots & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} \frac{\partial^2 \mathbf{u}}{\partial x_1 \partial x_1} & \frac{\partial^2 \mathbf{u}}{\partial x_2 \partial x_1} & \dots & \frac{\partial^2 \mathbf{u}}{\partial x_n \partial x_1} \\ \frac{\partial^2 \mathbf{u}}{\partial x_1 \partial x_2} & \ddots & & \vdots \\ \vdots & & \ddots & \vdots \\ \frac{\partial^2 \mathbf{u}}{\partial x_1 \partial x_n} & \dots & \dots & \frac{\partial^2 \mathbf{u}}{\partial x_n \partial x_n} \end{pmatrix} = A_L H^T.$$

Here the matrix H is a **Hessian**.

We are able to find the eigenvalues λ of the matrix A_L :

$$A_L \mathbf{v} = \lambda \mathbf{v}$$

- ▶ Suppose we have found several eigenvalues λ_i .
- ▶ **Elliptic equation:** $\lambda_i > 0$ or $\lambda_i < 0$ for $\forall i$.
All-positive or all-negative eigenvalues
- ▶ **Parabolic equation:** $\exists k : \lambda_k = 0$ while $\lambda_i > 0$ or $\lambda_i < 0$ for $\forall i \neq k$. There is only one zero eigenvalue while all others have the same sign.
- ▶ **Hyperbolic equation:** $\exists k : \lambda_k > 0$ while $\lambda_i < 0 \forall i \neq k$ OR
 $\exists k : \lambda_k > 0$ while $\lambda_i < 0 \forall i \neq k$
The signs of eigenvalues are the same excepting one of them, which is non-zero and has contrary sign.

Elliptic equations

The equation

$$L\mathbf{u} + F(\dots) = 0 \quad (1)$$

is considered elliptic if the coefficient matrix A_L has all-positive or all-negative eigenvalues.

- ▶ The solution is as smooth as the coefficients and boundary conditions allow.
- ▶ Well suited to describe static or equilibrium states.
- ▶ Less suitable for dynamic processes.

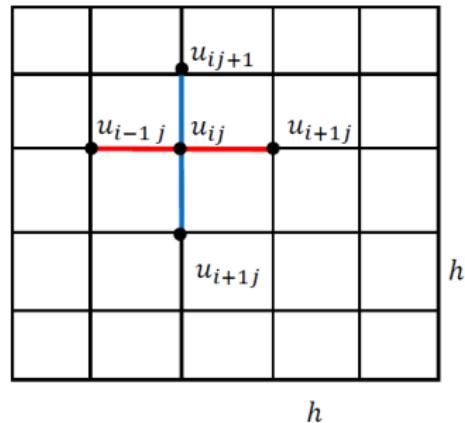
Meshed approaches on solution

- ▶ Finite Differences Method (FDM)
- ▶ Finite Elements Method (FEM)
- ▶ Finite Volumes Method (FVM)

FDM for elliptic problems: reminder

$$u_{xx}(x_i, y_j) \approx \frac{u_{i+1,j} - 2u_{ij} + u_{i-1,j}}{h_x^2}.$$

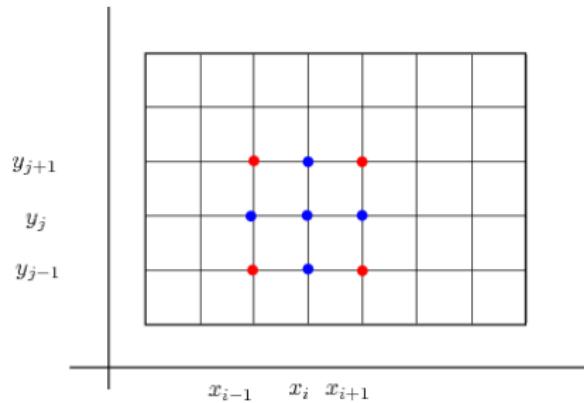
$$u_{yy}(x_i, y_j) \approx \frac{u_{i,j+1} - 2u_{ij} + u_{i,j-1}}{h_y^2}.$$



The Poisson equation in this approximation takes the form

$$-\left(\frac{u_{i+1,j} - 2u_{ij} + u_{i-1,j}}{h_x^2} + \frac{u_{i,j+1} - 2u_{ij} + u_{i,j-1}}{h_y^2} \right) = f_{ij}.$$

FDM: Mixed derivatives (2D case)

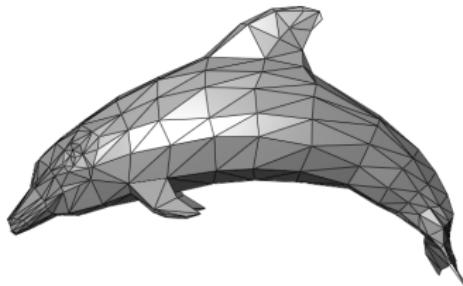


$$\frac{\partial^2 \mathbf{u}}{\partial x \partial y} \approx \frac{u_{i+1,j+1} - u_{i+1,j-1} - u_{i-1,j+1} + u_{i-1,j-1}}{4h_x h_y}$$

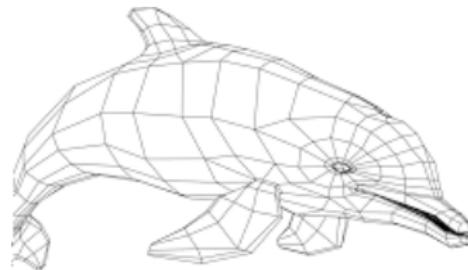
FDM: advantages and disadvantages

- ▶ Simplicity of mesh construction; simplicity of working with the mesh.
- ▶ In case of uniform meshes and simple boundary of the comp.domain, might work fast.
- ▶ In cases of static and slowly evolving systems, gives good accuracy.
- ▶ Can be too cumbersome for 3D meshes.
- ▶ High interpolation errors due to complex boundaries or complex inner structure.
- ▶ Evolving systems need to usage of implicit schemes, which affects on the performance.
- ▶ The system can be ill-conditioned.

FEM for elliptic problems: the mesh



$$r = 3, d = 3$$



$$r = 4, d = 3$$

- ▶ nodes: $N \times d$ array with d coordinates of N points in d -dimensional space.
- ▶ elements: $K \times r$ array with r indices of vertices in nodes array for K elements.

FEM for elliptic equation: remainder

- ▶ The equation: $Au = f, \quad u \in U.$
- ▶ Weak formulation:

$$(Au, v) = (f, v), \quad \forall v \in U.$$

- ▶ Approximation with FE:

$$u \approx \tilde{u} = \sum_{i=1}^N u_i \varphi_i(\mathbf{x}).$$

- ▶ Substitute \tilde{u} and take $v = \varphi_j, j = 1, \dots, N$. The SLAE with respect to u_i :

$$\sum_{i=1}^N u_i (A\varphi_i, \varphi_j) = (f, \varphi_j), \quad j = 1, \dots, N.$$

- ▶ Exclude u_i covered with boundary conditions.

- ▶ Suitable for computational domains with complex boundaries and/or complex inner structure.
- ▶ Not so cumbersome independently on the number of dimensions.
- ▶ Complicated procedure of mesh construction.
- ▶ The structure of the mesh may be complicated for working with it.
- ▶ The system is often being ill-conditioned.

The Neumann problem for Poisson equation in complex heterogeneous area

- ▶ The governing equation:

$$\nabla \cdot (\sigma(\mathbf{x}) \nabla u(\mathbf{x})) = \nabla \cdot J(\mathbf{x}), \quad \mathbf{x} \in \Omega \subset \mathbb{R}^3$$

- ▶ The Neumann boundary condition:

$$\frac{\partial u}{\partial n}|_{\partial\Omega} = 0.$$

- ▶ Continuity of the potential and its gradient on interfaces. Let $\partial\Omega_k$ be an intracranial interface. Denote by f^- and f^+ values of some function f nearby the interface, located on two sides of the boundary. Then

$$u^+|_{\partial\Omega_k} = u^-|_{\partial\Omega_k}; \quad \sigma^+ \frac{\partial u^+}{\partial n} = \sigma^- \frac{\partial u^-}{\partial n}.$$

The weak formulation of the problem

$$\int_{\Omega} (\sigma(\mathbf{x}) \nabla V(\mathbf{x})) \cdot \nabla h(\mathbf{x}) d\mathbf{x} = \int_{\Omega} (\nabla \cdot \mathbf{J}) h(\mathbf{x}) d\mathbf{x}.$$

In order to manage the discontinuous function \mathbf{J} , we avoid usage of its derivative in the RHS using the known integral identity:

$$\int_{\Omega} (\nabla \cdot \mathbf{J}(\mathbf{x})) h_m(\mathbf{x}) d\mathbf{x} + \int_{\Omega} \mathbf{J}(\mathbf{x}) \cdot \nabla h_m(\mathbf{x}) d\mathbf{x} = \int_{\partial\Omega} (\mathbf{J}(\mathbf{x}) \cdot \mathbf{n}) h_m(\mathbf{x}) d\mathbf{x}$$

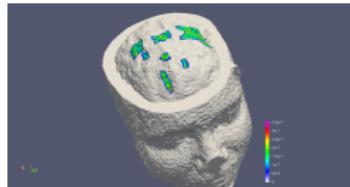
After discretization with linear finite elements we get the system to solve:

$$A u_h = b, \quad A_{nm} = - \int_{\Omega} \sigma(\mathbf{x}) \nabla h_m(\mathbf{x}) \cdot \nabla h_n(\mathbf{x}) d\mathbf{x},$$

$$b_m = \int_{\Omega} \nabla \cdot \mathbf{J}^i(\mathbf{x}) h_m(\mathbf{x}) d\mathbf{x} \equiv - \int_{\Omega} \mathbf{J}(\mathbf{x}) \cdot \nabla h_m(\mathbf{x})$$

The system properties and methods

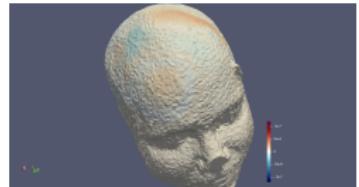
- ▶ The system contains $10^6 - 10^7$ equations and the same number of unknown variables;
- ▶ The matrix of the system is ill-conditioned;
- ▶ It's, however, symmetric and sparse;
- ▶ The suitable method to solve: generalized residual method with regularization.



$\mathbf{J}(\mathbf{x})$



$u(\mathbf{x})|_{\partial\Omega_2}$



$u(\mathbf{x})|_{\partial\Omega_1}$

Finite Volumes Method

Non-conservative difference scheme example

Consider the boundary value problem:

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

$$K(x) = \begin{cases} 1 & x < 1/2 \\ 5 & x > 1/2 \end{cases}$$

Using

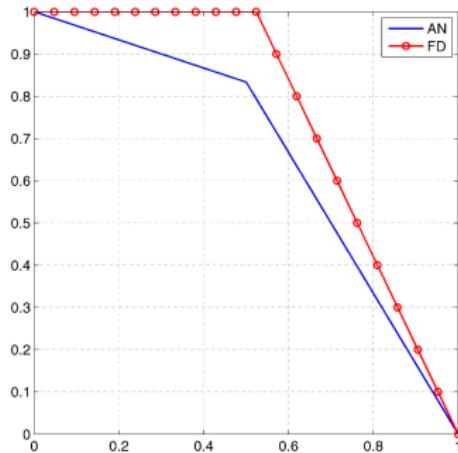
$$-Ku_{xx} - K_x u_x = 0,$$

build the difference scheme:

$$-K_j \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2} - \frac{K_{i+1} - K_{i-1}}{2h} \cdot \frac{u_{i+1} - u_{i-1}}{2h} = 0,$$

$$u_0 = 1, \quad u_n = 0$$

Non-conservative difference scheme example



The scheme does not converge!

Energy balance: conservative schemes

Consider the diffusion equation:

$$\begin{aligned}-\nabla \cdot (K \nabla U) &= f, \quad x \in \Omega, \\ u &= \varphi, \quad x \in \partial\Omega\end{aligned}$$

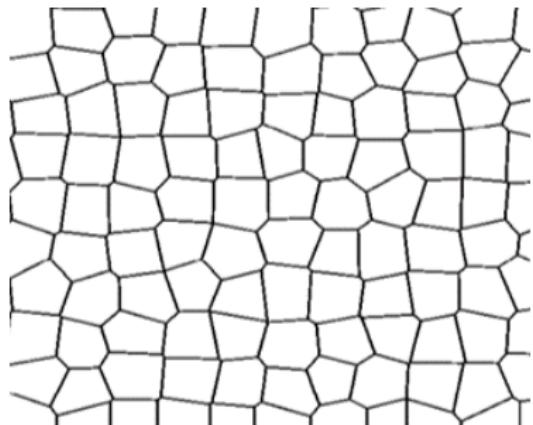
- ▶ $K(x)$ - the thermal conductivity
- ▶ \mathbf{W} - the thermal flux

$$\mathbf{W} = -K \nabla u,$$

$$\nabla \cdot \mathbf{W} = f$$

The energy balance equation

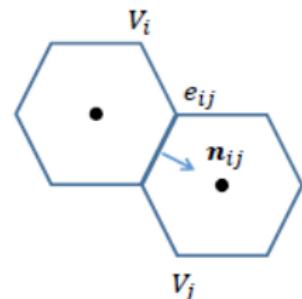
$$\int_{\partial V_i} \mathbf{W} \cdot \mathbf{n}_i dl = \int_{V_i} f dS$$



Finite Volumes Method

- ▶ e_{ij} - the link between the cell V_i and neighboring cell V_j
- ▶ \mathbf{n}_{ij} - the normal to e_{ij} with the direction to V_j
- ▶ W_{ij} - mean normal transfer through e_{ij}

$$W_{ij} = \frac{1}{|e_{ij}|} \int_{e_{ij}} \mathbf{W} \cdot \mathbf{n}_{ij} dl$$



- ▶
$$f_i = \frac{1}{|V_i|} \int_{V_i} f dS, u_i = \frac{1}{|V_i|} \int_{V_i} u dS$$

- ▶ Finally, the energy balance equation takes the form:

$$\sum_j |e_{ij}| W_{ij} = |V_i| f_i$$

Finite Volumes Method

- We need to express W_{ij} in terms of u_{ij} . Since $\mathbf{W} = -K \nabla u$:

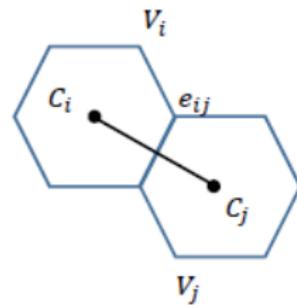
$$\frac{W_{ij}}{K} = -\nabla u \cdot \mathbf{n}_{ij}$$

- Let us integrate the latter identity between points C_i and C_j :

$$W_{ij} \int_{C_i C_j} \frac{1}{K} dl = - \int_{C_i C_j} \nabla u \cdot n_{ij} dl$$

- The length of interval $C_i C_j$ is h_{ij} and is divided into equal parts by the link e_{ij} . $K|_{V_i} = K_i$. Thus

$$W_{ij} = (u_i - u_j) \frac{2}{h_{ij}} \frac{K_i K_j}{K_i + K_j}$$



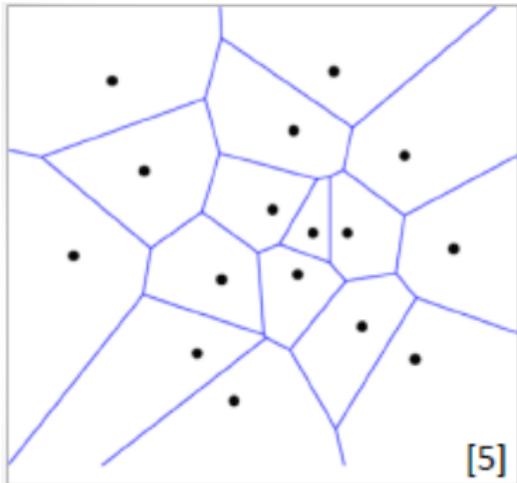
Thus, combining all reasoning above, we have:

$$\sum_j (u_i - u_j) \frac{|e_{ij}|}{h_{ij}} \frac{2K_i K_j}{K_i + K_j} = |v_i| f_i$$

The latter is SLAE with respect to u_i .

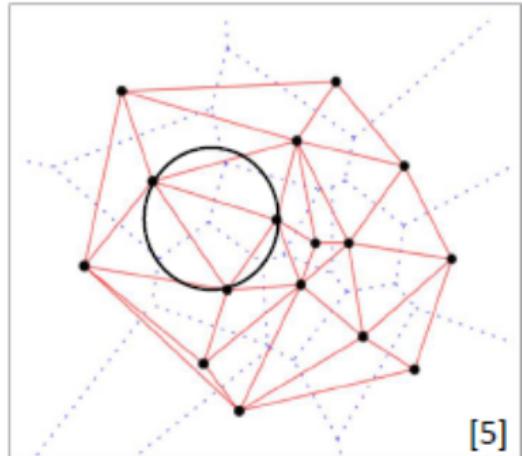
- ▶ In case of rectangular cells and constant K this SLAE is the same as Finite-difference SLAE
- ▶ The geometry is more flexible in comparison with FDM
- ▶ The accuracy: $O(h^2)$

Voronoi diagram



[5]

Let we have the set of point S . Each cell is consisting of all points closer to that seed than to any other.



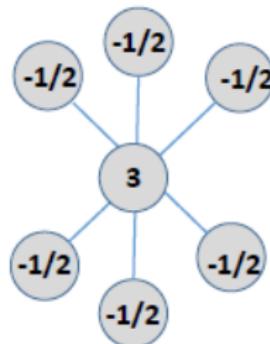
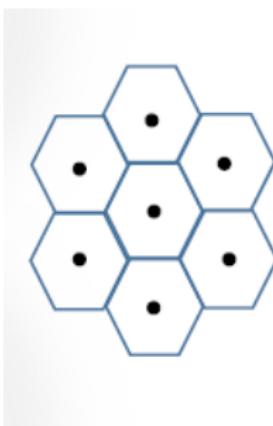
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The dual: Delaunay triangulation.

$$-\Delta u = f$$

Mesh constructed from regular hexagons

$$|e_{ij}| = \frac{h_{ij}}{2}, \quad K_{ij} = 1.$$

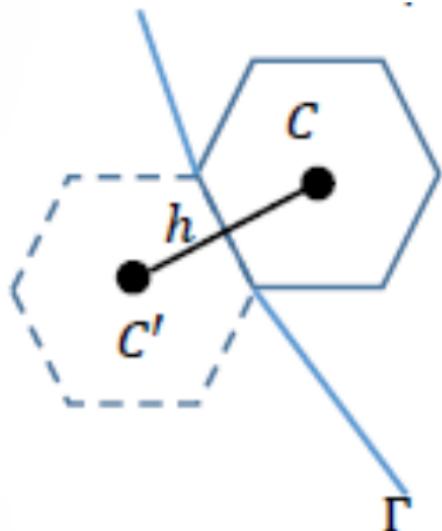


Approximation of boundary conditions

The Dirichlet boundary condition

$$u|_{\Gamma} = \varphi$$

can be taken into account using the fictitious cell:



$$\varphi = \frac{u_C + u_{C'}}{2} + O(h^2)$$

$$W_{CC'} = -K \frac{u_C - u_{C'}}{h}$$

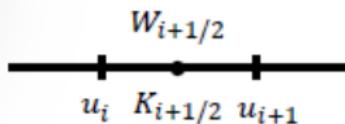
thus,

$$W_{CC'} = -K \frac{2u_C - 2\varphi}{h}$$

FVM templates

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

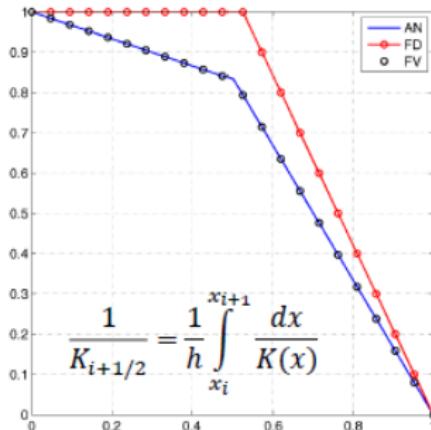
$$K(x) = \begin{cases} 1 & x < 1/2 \\ 5 & x > 1/2 \end{cases}, \quad W = -K(x)u_x, \quad W_x = 0.$$



Finite volume scheme:

$$W_{i+1/2} = -K_{i+1/2} \frac{u_{i+1} - u_i}{h}$$

$$\frac{W_{i+1/2} - W_{i-1/2}}{h} = 0$$



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Part 3: Parabolic equations
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The equation

$$L\mathbf{u} + F(\dots) = 0 \quad (2)$$

is considered parabolic if the coefficient matrix $\{a_{ij}\}$ has one zero eigenvalue, while all others have the same sign (positive or negative)

- ▶ May be represented in form $u_t = -L\mathbf{u}$, where \mathbf{u} is an elliptic operator.
- ▶ Are well suited to describe the smoothly evolving processes such diffusion or heat transfers.
- ▶ The solution is generally smoother than the initial value.

Parabolic equations: heat transfer equation

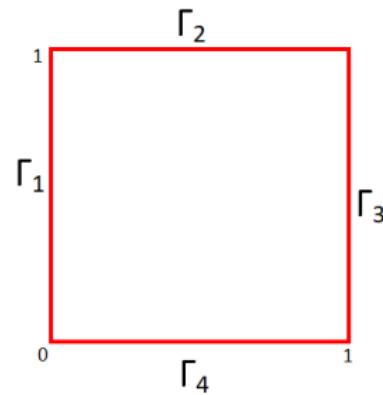
The quasilinear heat transfer equation:

$$u_t = \Delta u + aS(u) + F(\mathbf{x}, t),$$

$$(\mathbf{x}, t) \in [0, 1]^3,$$

$$u(\mathbf{x}, 0) = f(\mathbf{x}),$$

$$u_{\Gamma_k} = p_k(t), \quad k = 1, 2, 3, 4$$



The FDM implicit scheme

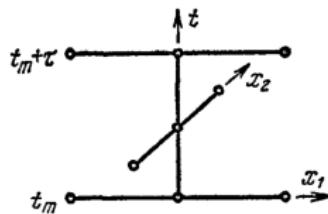
- ▶ Let τ be the time step.
- ▶ Consider three time layers: t , $\tilde{t} = t + \frac{\tau}{2}$, and $\hat{t} = t + \tau$.
- ▶ The layer \tilde{t} is named 'half-layer', and any function $\tilde{f} = f(\mathbf{x}, \tilde{t})$.
- ▶ The layer \hat{t} - the 'next' layer, and any function $\hat{f} = f(\mathbf{x}, \hat{t})$.
- ▶ Assume the values u_{nm} to be known in all nodes at t layer.
- ▶ Move from t to \tilde{t} via solution of SLAE:

$$\frac{2}{\tau}(\tilde{u}_{nm} - u_{nm}) = \Lambda_1 \tilde{u}_{nm} + \Lambda_2 u_{nm} + \tilde{f}_{nm}.$$

- ▶

$$\Lambda_1 u = \frac{u_{i+1,j} - 2u_{i,j} + u_{i-1,j}}{2h^2},$$

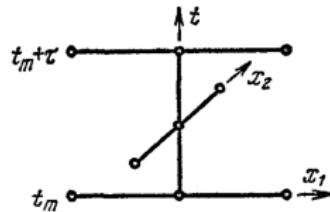
$$\Lambda_2 u = \frac{u_{i,j+1} - 2u_{i,j} + u_{i,j-1}}{2h^2}$$



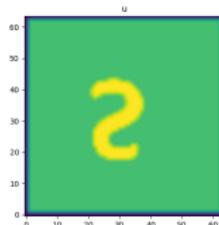
The FDM implicit scheme

- ▶ Consider three time layers: t , $\tilde{t} = t + \frac{\tau}{2}$, and $\hat{t} = t + \tau$.
- ▶ Assume the values u_{nm} to be known in all nodes at t layer.
- ▶ Move from t to \tilde{t} .
- ▶ Move from \tilde{t} to \hat{t} :

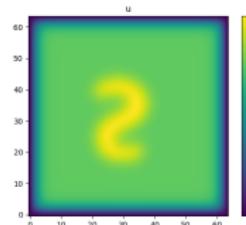
$$\frac{2}{\tau}(\hat{u}_{nm} - \tilde{u}_{nm}) = \Lambda_1 \tilde{u}_{nm} + \Lambda_2 \hat{u}_{nm} + \tilde{f}_{nm}.$$



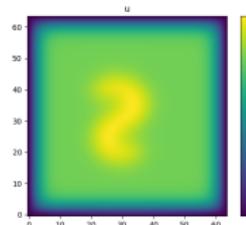
2D Parabolic: result with FDM



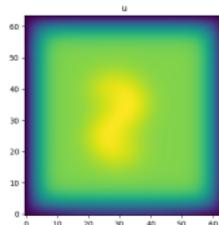
$t = 0$



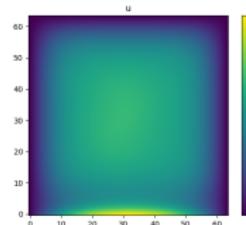
$t = 10s$



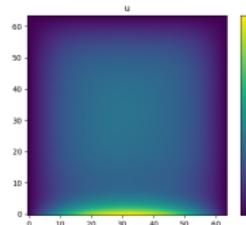
$t = 20s$



$t = 40$



$t = 100s$



$t = 200s$

2D heat transfer: solution using FEM

- ▶ The weak formulation:

$$\int_{\Omega} p du - f tv dx + \int_{\Omega} \nabla u \cdot \nabla v dx - \int_{\Omega} cuv dx$$

- ▶ The software to use: Fenics/dolfin (C++/python), OpenFoam.

The ill-posed Cauchy problem for heat transfer

The ill-posed Cauchy problem for heat transfer equation.

Suppose both initial condition $f(x)$ and the left boundary condition $g(t)$ to be unknown, and the function $q(t) = \partial u / \partial x(1, t)$ to be known:

$$u_t = u_{xx} + aS(u) + F(x, t), \quad (x, t) \in [0, 1]^2,$$
$$u(1, t) = p(t), \quad \frac{\partial u}{\partial x}(1, t) = q(t).$$

Find $u(x, t)$.

The problem is non-linear (quasi-linear), which leads that the cost functional has local minima

Coefficient problem for a heat-transfer equation

Let $\Omega \subset \mathbb{R}^n$ be some domain, $\partial\Omega$ - its boundary, and $\Gamma \subset \partial\Omega$ is some part of this boundary. Consider the problem:

$$u_t = \Delta u + a(\mathbf{x})u, \quad \mathbf{x} \in \Omega$$

$$u_\Gamma(\mathbf{x}, t) = p(\mathbf{x}, t).$$

Assuming $u(\mathbf{x}) > 0, \mathbf{x} \in \Omega$ introduce the notation: $u = e^\nu$, where $\nu(x, t)$ is some function.

$$\nu_t u = u \Delta \nu + u (\nabla \nu)^2 + a(\mathbf{x})u, \quad \nu|_\Gamma = \ln(p).$$

Dividing the equation by u , we obtain:

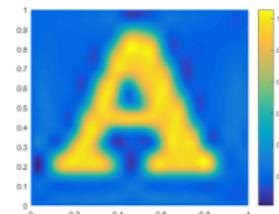
$$\nu_t = \Delta \nu + (\nabla \nu)^2 + a(\mathbf{x}).$$

Now, differentiating the latter equation on t , we just exclude the coefficient $a(x)!$

Heat transfer CIP results



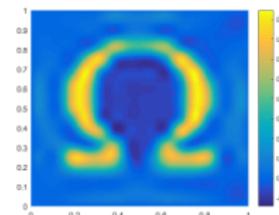
(a) Coefficient with shape 'A'



(b) Recovered $a(x,y)$



(c) Coefficient with shape ' Ω '



(d) Recovered $a(x,y)$

The result was obtained by Prof. Klibanov (UNCC Charlotte) with co-authors.

Scientific Computing

Lecture 5

Part 2: Hyperbolic equations
Nikolay Koshev, Nikolay Yavich

October 14, 2021

Skoltech

Skolkovo Institute of Science and Technology

The equation

$$L\mathbf{u} + F(\dots) = 0 \quad (3)$$

is considered hyperbolic if the coefficient matrix $\{a_{ij}\}$ has eigenvalues, the signs of which are the same excepting one of them, which is non-zero and has contrary sign.

- ▶ May be represented with elliptic operator:
 $L\mathbf{u} - a^2 \frac{\partial^2 \mathbf{u}}{\partial t^2} = F(\dots)$, where L is an elliptic operator.
- ▶ Hyperbolic equations solutions retain discontinuities of initial data.
- ▶ Well-suited to describe wave processes.

The Maxwell system

- ▶ The Gauss Law (in the sourceless domain)

$$\nabla \cdot \mathbf{E} = 0$$

- ▶ The Gauss Law for Magnetic induction

$$\nabla \cdot \mathbf{B} = 0$$

- ▶ The Faraday's Law

$$\nabla \times \mathbf{E} = \frac{\partial \mathbf{B}}{\partial t}$$

- ▶ The Ampere's Circuital Law

$$\nabla \times \mathbf{B} = \mu_0 \varepsilon_0 \frac{\partial \mathbf{E}}{\partial t}$$

The Wave equation

- ▶ Applying curl to the Faraday's law:

$$\nabla \times (\nabla \times \mathbf{E}) = -\mu_0 \epsilon_0 \frac{\partial^2 \mathbf{E}}{\partial t^2}$$

- ▶ Applying curl to the Ampere's law:

$$\nabla \times (\nabla \times \mathbf{B}) = -\mu_0 \epsilon_0 \frac{\partial^2 \mathbf{B}}{\partial t^2}$$

- ▶ Using the identity $\nabla \times \nabla \times \mathbf{E} = \nabla(\nabla \cdot \mathbf{E}) - \Delta \mathbf{E}$:

$$\frac{1}{c^2} \frac{\partial^2 \mathbf{E}}{\partial t^2} - \Delta \mathbf{E} = 0$$

$$\frac{1}{c^2} \frac{\partial^2 \mathbf{B}}{\partial t^2} - \Delta \mathbf{B} = 0.$$

The weak form

Consider 1D boundary value problem:

$$\frac{\partial^2 u}{\partial t^2} = k \frac{\partial^2 u}{\partial x^2}, \quad x \in [a, b], t \in [0, T]$$

$$u(x, 0) = 0, \frac{\partial u}{\partial t}(x, 0) = V_0(x), u(a, t) = 0, k \frac{\partial u}{\partial x}(b, t) = t_b(t)$$

- ▶ These conditions make the problem well-defined.
- ▶ Integrating it:

$$\int_a^b v \frac{\partial^2 u}{\partial t^2} dx = k \int_a^b v \frac{\partial^2 u}{\partial x^2} dx$$

The weak form

Consider 1D boundary value problem:

$$\frac{\partial^2 u}{\partial t^2} = k \frac{\partial^2 u}{\partial x^2}, \quad x \in [a, b], t \in [0, T]$$

$$u(x, 0) = 0, \frac{\partial u}{\partial t}(x, 0) = V_0(x), u(a, t) = 0, k \frac{\partial u}{\partial x}(b, t) = t_b(t)$$

- ▶ Integrating the right-hand side by parts, we get:

$$\int_a^b v \frac{\partial^2 u}{\partial t^2} dx = vt_b(t) - \int_a^b \frac{\partial v}{\partial x} \frac{\partial u}{\partial x} (x, t) dx$$

- ▶ In nD case:

$$\int_{\Omega} \frac{\partial^2 u}{\partial t^2} v d\mathbf{x} = \Phi(\partial\Omega) - k \int_{\Omega} \nabla u \cdot \nabla v d\mathbf{x}$$

The methods: FEM

- Finite elements:

$$u(x, t) = \sum_{i=1}^N u_i(t) \varphi_i(x)$$

- Substituting it to the weak form, we get the system:

$$\mathbf{M}\ddot{\mathbf{u}}(t) + \mathbf{K}\mathbf{u}(t) - \mathbf{t}_b(t) = 0,$$

$$M_{ij} = \int_a^b \varphi_i(x) \varphi_j(x) dx, \quad K_{ij} = \int_a^b \frac{\partial \varphi_i}{\partial x} \cdot \frac{\partial \varphi_j}{\partial x}, \\ i, j = 1, 2, \dots, N.$$

Backward Euler

- ▶ The system

$$\mathbf{M}\ddot{\mathbf{u}}(t) + \mathbf{Ku}(t) - t_b(t) = 0,$$

- ▶ Rewrite it in the form:

$$\frac{\partial}{\partial t} \begin{pmatrix} \mathbf{u}(t) \\ \mathbf{M}\dot{\mathbf{u}}(t) \end{pmatrix} = \begin{pmatrix} \dot{\mathbf{u}}(t) \\ -\mathbf{Ku}(t) + t_b(t) \end{pmatrix}$$

- ▶ Discretization on t (backward Euler scheme, $\mathbf{v} = \dot{\mathbf{u}}$):

$$\begin{pmatrix} \frac{\mathbf{u}^{n+1} - \mathbf{u}^n}{h_t} \\ \mathbf{M} \left(\frac{\mathbf{v}^{n+1} - \mathbf{v}^n}{h_t} \right) \end{pmatrix} = \begin{pmatrix} \mathbf{v}^{n+1} \\ -\mathbf{Ku}^{n+1} + t_b(t^{n+1}) \end{pmatrix}$$

The FDM discretization

Consider the BV problem:

$$u_{tt} = c^2 u_{xx} + f(x, t), \quad 0 < x < a, 0 < t \leq T$$

$$u(x, 0) = \mu_1(x), \quad u_t(x, 0) = \mu_2(x), \quad 0 < x < a$$

$$u(0, t) = \mu_3(t), \quad u(a, t) = \mu_4(t), \quad 0 \leq t \leq T.$$

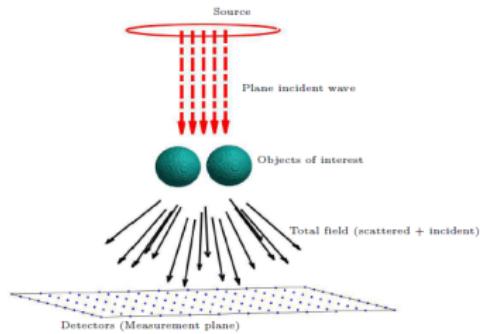
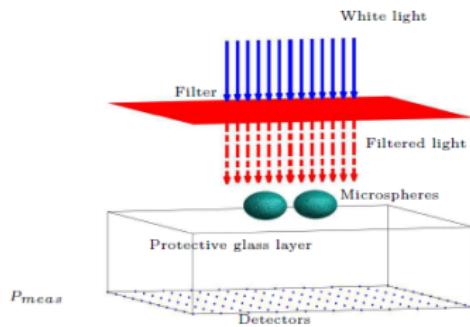
► The Cross template:

$$\frac{1}{\tau^2}(\hat{y}_n - 2y_n + \check{y}_n) = \frac{c^2}{h^2}(y_{n+1} - 2y_n + y_{n-1}) + f_n.$$

- On the initial and the next layers the solution is known from initial conditions
- The scheme is explicit and allows to get \hat{y}_n from known y_n, \check{y}_n .

The Plane Wave Diffraction

The microscpheres are being irradiated with a plane wave of visible-light frequencies



Source: *A numerical method to solve a phaseless coefficient inverse problem from a single measurement of experimental data.* MV Klibanov, NA Koshev, DL Nguyen, LH Nguyen, A Bretin, VN Astratov *SIAM Journal on Imaging Sciences*, Vol. 11 (4), 2339-2367, 2018

Wave equation to Helmholtz equation

- ▶ The equation for the electric field:

$$\frac{1}{c^2} \frac{\partial^2 \mathbf{E}}{\partial t^2} - \Delta \mathbf{E} = 0. \quad (4)$$

- ▶ Since \mathbf{E} is a plane wave, then $\mathbf{E} = u(x) \exp(i\omega t)$
- ▶ Substituting it to the equation, we get

$$-\frac{\omega^2}{c^2} u(x) \exp(i\omega t) - \Delta u(x) \exp(i\omega t) = 0$$

- ▶ Dividing by $-\exp(...)$ and denoting $\omega^2/c^2 = k^2$:

$$\Delta u(\mathbf{x}) + k^2 u(\mathbf{x}, k) = 0.$$

- ▶ The obtained equation is a Helmholtz equation; and it's elliptic!

Integral representation example: the Helmholtz equation

Consider the plane wave propagating in 3D. The governing equation is the Helmholtz equation:

$$\begin{aligned}\Delta u(\mathbf{x}, k) + k^2 n^2(\mathbf{x}) u(\mathbf{x}, k) &= 0, \quad \mathbf{x} \in \mathbb{R}^3 \\ \partial_{|\mathbf{x}|} u_{sc}(\mathbf{x}, k) - iku_{sc}(\mathbf{x}, k) &= o(|\mathbf{x}|^{-1}) \quad |\mathbf{x}| \rightarrow \infty.\end{aligned}$$

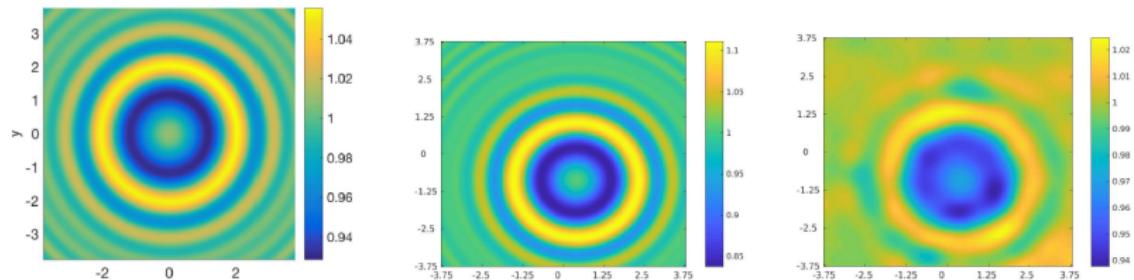
The problem above is called **Scattering problem**.

This is an equivalent for it: the **Lippmann-Schwinger equation**:

$$\begin{aligned}u(\mathbf{x}, k) &= e^{ikx_3} + k^2 \int_{\Omega} \frac{\exp(ik|\mathbf{x} - \xi|)}{3\pi|\mathbf{x} - \xi|} (n^2(\xi) - 1) u(\xi, k) d\xi. \\ v(\mathbf{x}, k) g(\mathbf{x}) &= e^{ikx_3} + k^2 K * v, \quad K(\mathbf{x}) = \frac{\exp(ik|\mathbf{x}|)}{3\pi|\mathbf{x}|}, v = (n^2(\mathbf{x}) - 1) u(\mathbf{x}, k).\end{aligned}$$

QUESTION: Classify the IE above.

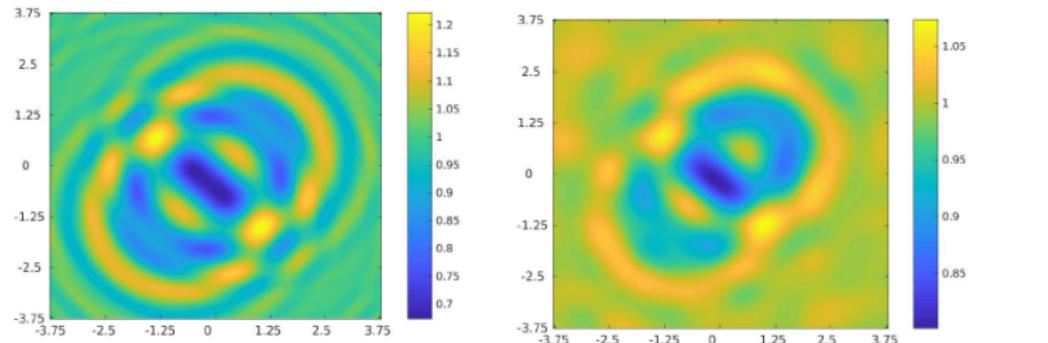
The Plane Wave Diffraction



Simulation (MS)

Simulation (LSE, HG)

Experimental

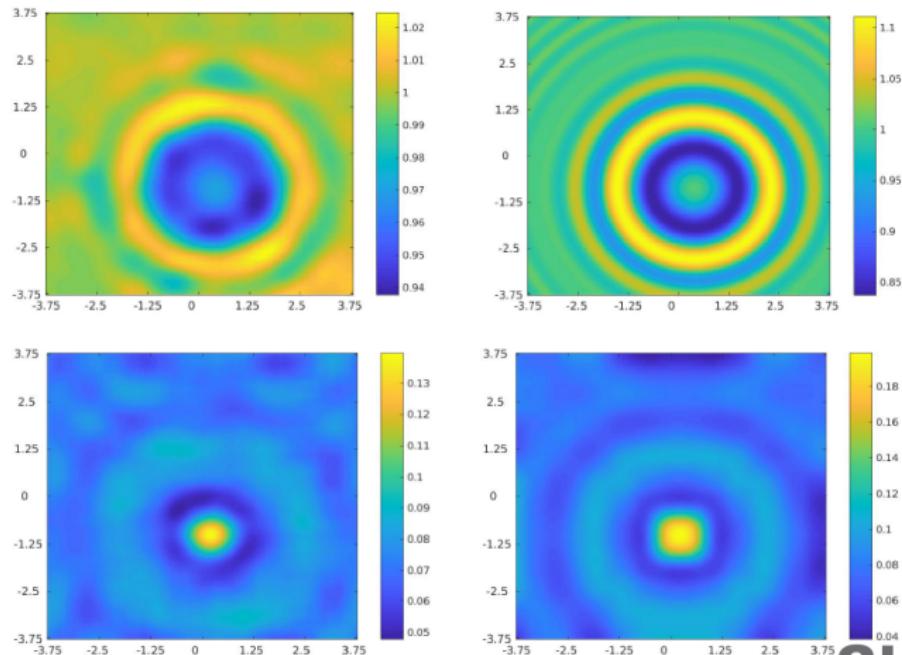


Simulation (LSE, HG)

Experimental

The Backpropagation

By the backpropagation we mean the backwards (inverse) solution of the Lippmann-Schwinger equation in order to 'focus' the data near the protective glass layer (far away from the sensors).



Scientific Computing

Lecture 5

Part 5: Cauchy problem for Elliptic equation example
Nikolay Koshev, Nikolay Yavich

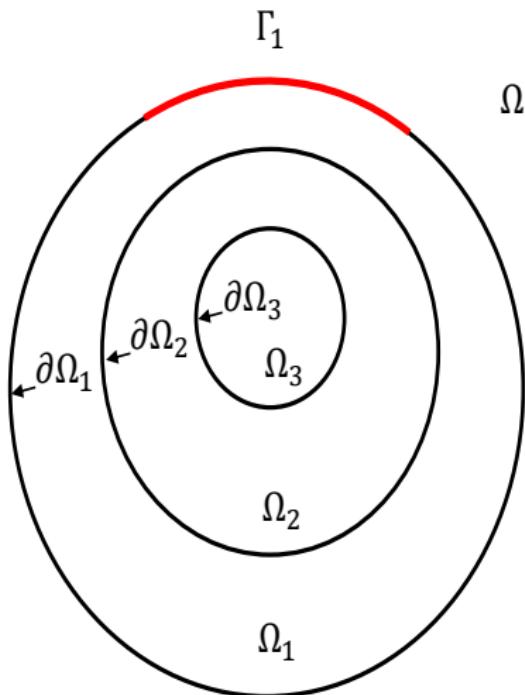
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Source localization with data mapping

The EEG Cauchy problem (Data mapping): the domain



► Notations:

- ▶ $\Omega \subset \mathbb{R}^3$ - the whole domain.
- ▶ $\Omega_k, k = 1, \dots, N_\Omega - 1$ - the outer 'sourceless' domains ($\sigma(\Omega_k) = \sigma_k = \text{const}$).
- ▶ Ω_{N_Ω} - the inner domain containing sources ($\sigma(\Omega_{N_\Omega}) = \sigma_{N_\Omega} = \text{const}$)
- ▶ The accessible part of the boundary

$$\Gamma_k = \begin{cases} \Gamma_1, & k = 1 \\ \partial\Omega_k, & k \geq 2. \end{cases}$$

- ▶ The rest of the boundary: Π_k .

► Assumptions:

- ▶ Current density volumetric distribution is non-zero only in the inner domain:
 $\mathbf{J} : \text{supp}(\mathbf{J}) \in \Omega_{N_\Omega}$
- ▶ Conductivities:
 $\sigma_k \ll \sigma_{N_\Omega}, k = 1, \dots, N_\Omega - 1$.

Data mapping: the Cauchy problem

Assume the function $u(\mathbf{x})$, $\mathbf{x} \in \Pi_1 \subset \partial\Omega_1$ to be known. Find the function $U(\mathbf{x})$, $\mathbf{x} \in \Omega_1$, such that:

$$\Delta U = 0, \quad \mathbf{x} \in \Omega_1;$$

$$U(\mathbf{x}) = u(\mathbf{x}), \quad \mathbf{x} \in \Gamma_1;$$

$$\frac{\partial U}{\partial \mathbf{n}}(\mathbf{x}) = g(\mathbf{x}), \quad \mathbf{x} \in \Gamma_1,$$

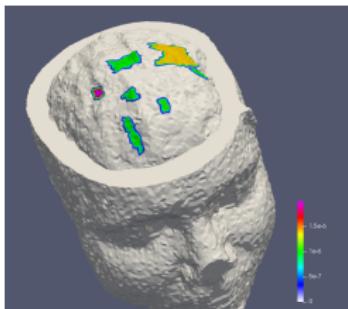
where

$$g(\mathbf{x}) = \begin{cases} 0, & \Gamma_1 \subset \partial\Omega, \\ \frac{\sigma_1}{\sigma_2} \mathbf{n} \cdot \nabla U(\mathbf{x}), & \Gamma_1 \subset \partial\Omega_2. \end{cases}$$

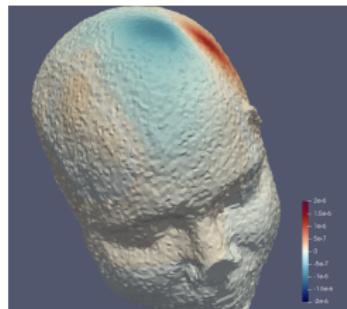
Whenever the function $U(\mathbf{x})$ is found at every point $\mathbf{x} \in \Omega_1$, we obviously have the potential on the rest of a boundary

$$\Pi_1 = \partial\Omega_1 \setminus \Gamma_1.$$

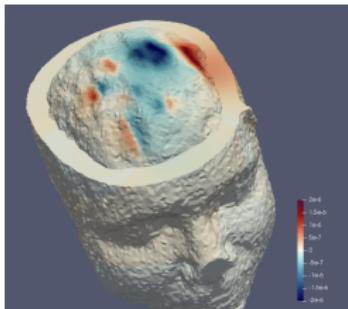
Numerical results: regular currents



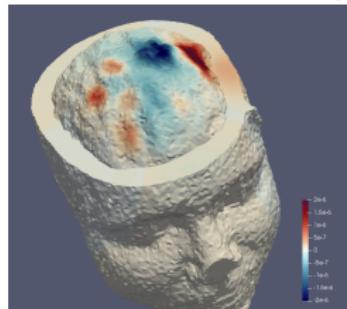
The sources



Potential on the scalp

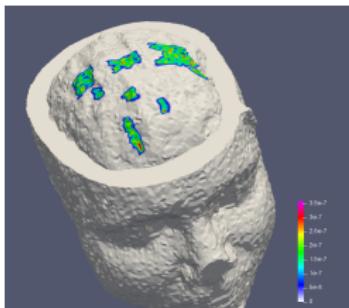


Real potential (brain)

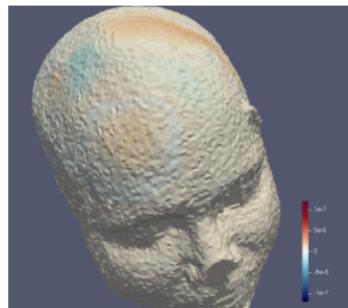


Reconstruction result

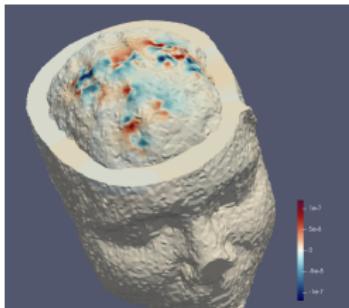
Numerical results: Random currents in each point



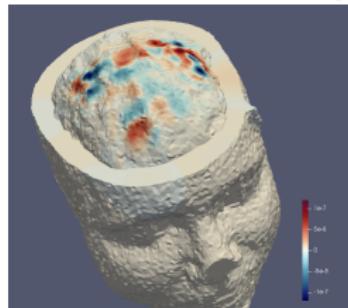
The sources



Potential on the scalp



Real potential (brain)



Reconstruction result

The algorithm. Let

- ▶ Create the FEM mesh allowing to assign appropriate number N_d of subdomains $\Theta_i, i = 1, \dots, N_d$
- ▶ Solve the Cauchy problem (??) for $i = 1$ in order to obtain the function U_1
- ▶ For $i = 2, \dots, N_d$ do the following:
 - ▶ Calculate U_i solving the Cauchy problem (??)
 - ▶ Calculate the number $e_i = \|U_{i-1} - U_i\|_{L^2(\Theta_{i-1})}$
- ▶ Find the index i in which $e_i - e_{i-1} > l$, where $l > 0$ is some number, chosen, for example, as a mean value over e ;
- ▶ The index $i \equiv s$ due to the consideration above.

The results

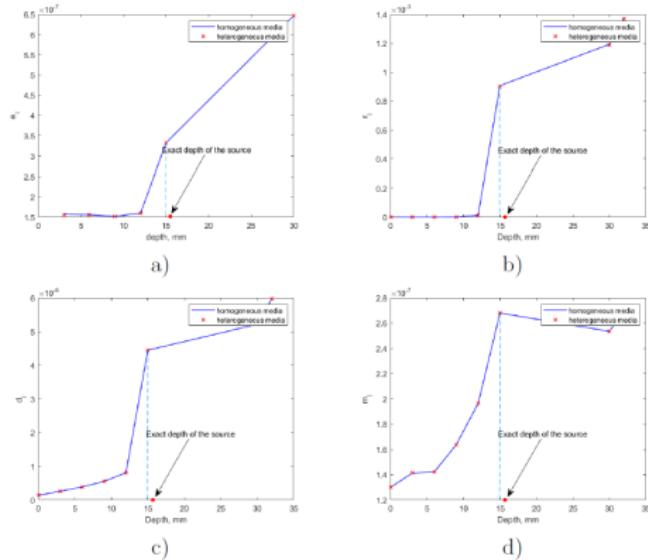


Figure 7: Dependencies on the depth of: a) e_i , b) - relative residual r_i ; c) norm of Laplacian of the numerical solution l_i , and d) maximum absolute values of the potential m_i