

# Numerical software 2

## Anisotropic mesh adaptation

V. Dolejší

Charles University Prague, Faculty of Mathematics and Physics

Lecture 1

## Numerical solution of PDE

- we seek  $u : \Omega \rightarrow \mathbb{R}$  such that  $\mathcal{L}u = f$  in  $\Omega$
- we define mesh  $\mathcal{T}_h$  of  $\Omega$  and finite dimensional space  $V_h$
- approximate solution  $u_h \in V_h$

## Main goal

Define (create) a mesh  $\mathcal{T}_h$  such that

- 1 the computational error is under the given tolerance
- 2 the number of elements of  $\mathcal{T}_h$  is as small as possible

## Fundamental question

How to fulfil the main goal?

- we adapt the given mesh based on the computed solution and its error estimation

## Numerical solution of PDE

- we seek  $u : \Omega \rightarrow \mathbb{R}$  such that  $\mathcal{L}u = f$  in  $\Omega$
- we define mesh  $\mathcal{T}_h$  of  $\Omega$  and finite dimensional space  $V_h$
- approximate solution  $u_h \in V_h$

## Main goal

Define (create) a mesh  $\mathcal{T}_h$  such that

- 1 the computational error is under the given tolerance
- 2 the number of elements of  $\mathcal{T}_h$  is as small as possible

## Fundamental question

How to fulfil the main goal?

- we adapt the given mesh based on the computed solution and its error estimation

## Numerical solution of PDE

- we seek  $u : \Omega \rightarrow \mathbb{R}$  such that  $\mathcal{L}u = f$  in  $\Omega$
- we define mesh  $\mathcal{T}_h$  of  $\Omega$  and finite dimensional space  $V_h$
- approximate solution  $u_h \in V_h$

## Main goal

Define (create) a mesh  $\mathcal{T}_h$  such that

- 1 the computational error is under the given tolerance
- 2 the number of elements of  $\mathcal{T}_h$  is as small as possible

## Fundamental question

How to fulfil the main goal?

- we adapt the given mesh based on the computed solution and its error estimation

## Main idea

- let  $u$  be the exact solution and  $u_h \in V_h$  the approximate one
- let  $\Pi_h : V \rightarrow V_h$  be a projection
- we approximate  $u - u_h \approx u - \Pi_h u$
- $u - \Pi_h u =$  interpolation error

## Formulation of an abstract problem

- Let  $u : \Omega \rightarrow \mathbb{R}$  be a given function and  $\omega > 0$
- let  $\Pi_h : V \rightarrow V_h$  be a projection
- We seek  $\mathcal{T}_h$  such that
  - 1  $\|u - \Pi_h\| \leq \omega$
  - 2  $\#\mathcal{T}_h$  be minimal

exact solution  $u$  is unknown, it will be later approximated by  $u_h$

## Main idea

- let  $u$  be the exact solution and  $u_h \in V_h$  the approximate one
- let  $\Pi_h : V \rightarrow V_h$  be a projection
- we approximate  $u - u_h \approx u - \Pi_h u$
- $u - \Pi_h u =$  interpolation error

## Formulation of an abstract problem

- Let  $u : \Omega \rightarrow \mathbb{R}$  be a given function and  $\omega > 0$
- let  $\Pi_h : V \rightarrow V_h$  be a projection
- We seek  $\mathcal{T}_h$  such that
  - 1  $\|u - \Pi_h\| \leq \omega$
  - 2  $\#\mathcal{T}_h$  be minimal

exact solution  $u$  is unknown, it will be later approximated by  $u_h$

## Main idea

- let  $u$  be the exact solution and  $u_h \in V_h$  the approximate one
- let  $\Pi_h : V \rightarrow V_h$  be a projection
- we approximate  $u - u_h \approx u - \Pi_h u$
- $u - \Pi_h u =$  interpolation error

## Formulation of an abstract problem

- Let  $u : \Omega \rightarrow \mathbb{R}$  be a given function and  $\omega > 0$
- let  $\Pi_h : V \rightarrow V_h$  be a projection
- We seek  $\mathcal{T}_h$  such that
  - 1  $\|u - \Pi_h\| \leq \omega$
  - 2  $\#\mathcal{T}_h$  be minimal

exact solution  $u$  is unknown, it will be later approximated by  $u_h$

- $V_h = \{v_h \in L^2(\Omega); v_h|_K \in P^1(K) \forall K \in \mathcal{T}_h\}$  – discontinuous piecewise linear
- $\|\cdot\| := \|\cdot\|_{L^\infty(\Omega)}$
- $\Pi_h : V \rightarrow V_h$  such that
  - 1  $\Pi_h u(x_K) = u(x_K)$ ,  $x_K$  is the barycentre of  $K \in \mathcal{T}_h$
  - 2  $\nabla \Pi_h u(x_K) = \nabla u(x_K)$ ,  $x_K$  is the barycentre of  $K \in \mathcal{T}_h$

$\Pi_h u$  is a discontinuous piecewise linear, the same value and gradient as  $u$  in the barycentres of all  $K \in \mathcal{T}_h$

## Interpolation error

$$u(x) = \underbrace{u(x_K) + \nabla u(x_K)(x - x_K)}_{\Pi_h u(x)} + \frac{1}{2}(x - x_K)^T \mathcal{H}(u(x'))(x - x_K)$$

$\mathcal{H}(u(\cdot))$  ... Hessian matrix



- $V_h = \{v_h \in L^2(\Omega); v_h|_K \in P^1(K) \forall K \in \mathcal{T}_h\}$  – discontinuous piecewise linear
- $\|\cdot\| := \|\cdot\|_{L^\infty(\Omega)}$
- $\Pi_h : V \rightarrow V_h$  such that
  - 1  $\Pi_h u(x_K) = u(x_K)$ ,  $x_K$  is the barycentre of  $K \in \mathcal{T}_h$
  - 2  $\nabla \Pi_h u(x_K) = \nabla u(x_K)$ ,  $x_K$  is the barycentre of  $K \in \mathcal{T}_h$

$\Pi_h u$  is a discontinuous piecewise linear, the same value and gradient as  $u$  in the barycentres of all  $K \in \mathcal{T}_h$

## Interpolation error

$$u(x) = \underbrace{u(x_K) + \nabla u(x_K)(x - x_K)}_{\Pi_h u(x)} + \frac{1}{2}(x - x_K)^T \mathcal{H}(u(x'))(x - x_K)$$

$\mathcal{H}(u(\cdot))$  ... Hessian matrix

# Interpolation error estimate

$$u(x) - \Pi_h u(x) \approx \frac{1}{2}(x - x_K)^T \mathcal{H}(u(x_K))(x - x_K),$$

$$\mathcal{H}(u(x)) = \begin{pmatrix} \frac{\partial^2 u(x)}{\partial x_1^2} & \frac{\partial^2 u(x)}{\partial x_1 \partial x_2} \\ \frac{\partial^2 u(x)}{\partial x_1 \partial x_2} & \frac{\partial^2 u(x)}{\partial x_2^2} \end{pmatrix}$$

we assume that  $\mathcal{H}(u)$  is positively definite then  
(up to a higher order terms)

## Interpolation error estimate

$$|u(x) - \Pi_h u(x)| \leq \frac{1}{2}(x - x_K)^T \mathcal{H}(u(x_K))(x - x_K),$$

## Interpolation error estimate

$$|u(x) - \Pi_h u(x)| \leq \frac{1}{2}(x - x_K)^T \mathcal{H}(u(x_K))(x - x_K), \quad x \in K$$

## Our goal

$$\|u(x) - \Pi_h u(x)\| \leq \omega \quad \Leftrightarrow \quad |u(x) - \Pi_h u(x)| \leq \omega \quad \forall x \in K$$

## Equivalent condition

$$\frac{1}{2}(x - x_K)^T \mathcal{H}(u(x_K))(x - x_K) \leq \omega \quad \forall x \in K \quad (1)$$

## Geometrical interpretation

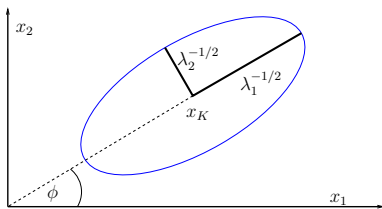
all  $x \in \mathbb{R}^2$  satisfying (1) form an ellipse

- let  $\mathbf{M}$  be a symmetric, positively definite, then

$$\mathbf{M} = \mathbf{R}^T \mathbf{L} \mathbf{R},$$

where  $\mathbf{R} = \begin{pmatrix} \cos \phi & \sin \phi \\ -\sin \phi & \cos \phi \end{pmatrix}$ ,  $\mathbf{L} = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$ ,

$$(x - x_K)^T \mathbf{M} (x - x_K) \leq 1 \iff x \in E$$



# Fulfilling of requirements

let  $\mathcal{H}(u(x_K))$  be the Hessian,  $E_K$  be the corresponding ellipse

## Lemma

$\frac{1}{2}(x - x_K)^T \mathcal{H}(u(x_K))(x - x_K) \leq \omega \quad \forall x \in K$  is valid  $\Leftrightarrow K \subset E_K$

## Definition

$K$  is **optimal triangle**  $\Leftrightarrow K \subset E_K$  & area of  $K$  is maximal

## Lemma

Let  $\mathcal{H} = \mathcal{H}(u(x_K))$  be the Hessian,  $E_K$  the ellipse.

Then  $K$  is **optimal triangle**  $\Leftrightarrow$

$$\|e_{K,i}\|_{\mathcal{H}} := \left( e_{K,i}^T \mathcal{H} e_{K,i} \right)^{1/2} = \sqrt{6\omega}, \quad i = 1, 2, 3,$$

where  $e_{K,i}$ ,  $i = 1, 2, 3$  are edges of  $K$ .

# Edge-based optimality

- Let  $K$  and  $K'$  share edge  $e$ ,  $\mathcal{H}$  and  $\mathcal{H}'$  be the Hessians
- then  $\|e\|_{\mathcal{H}} = \sqrt{6\omega} = \|e\|_{\mathcal{H}'}$  **can not be valid!!**
- it is impossible to consider  $\mathcal{T}_h$  as a set of “optimal triangles”

## Edge-based mesh

- Let  $\mathcal{F}_h$  denotes the set of edges  $e$  of the mesh  $\mathcal{T}_h$
- Let  $\mathcal{H}_e$  be the Hessian evaluated at edge  $e \in \mathcal{F}_h$ .
- Def: **mesh  $\mathcal{T}_h$  is edge-optimal**  $\Leftrightarrow \|e\|_{\mathcal{H}_e} = \sqrt{6\omega} \quad \forall e \in \mathcal{F}_h$

Edge optimal mesh exists only in special situation

## Definition

Mesh  $\mathcal{T}_h$  is optimal  $\Leftrightarrow Q_{\mathcal{T}_h} = \min_{\mathcal{T}_{h'}} Q_{\mathcal{T}_{h'}}$

where  $Q_{\mathcal{T}_{h'}} = \frac{1}{\#\mathcal{F}_{h'}} \sum_{e \in \mathcal{F}_{h'}} (\|e\|_{\mathcal{H}_e} - \sqrt{6\omega})^2$

$Q_{\mathcal{T}_h} \geq 0$  ... parameter of “quality”

# Edge-based optimality

- Let  $K$  and  $K'$  share edge  $e$ ,  $\mathcal{H}$  and  $\mathcal{H}'$  be the Hessians
- then  $\|e\|_{\mathcal{H}} = \sqrt{6\omega} = \|e\|_{\mathcal{H}'}$  **can not be valid!!**
- it is impossible to consider  $\mathcal{T}_h$  as a set of “optimal triangles”

## Edge-based mesh

- Let  $\mathcal{F}_h$  denotes the set of edges  $e$  of the mesh  $\mathcal{T}_h$
- Let  $\mathcal{H}_e$  be the Hessian evaluated at edge  $e \in \mathcal{F}_h$ .
- Def: **mesh  $\mathcal{T}_h$  is edge-optimal**  $\Leftrightarrow \|e\|_{\mathcal{H}_e} = \sqrt{6\omega} \quad \forall e \in \mathcal{F}_h$

Edge optimal mesh exists only in special situation

## Definition

Mesh  $\mathcal{T}_h$  is optimal  $\Leftrightarrow Q_{\mathcal{T}_h} = \min_{\mathcal{T}_{h'}} Q_{\mathcal{T}_{h'}}$

where  $Q_{\mathcal{T}_{h'}} = \frac{1}{\#\mathcal{F}_{h'}} \sum_{e \in \mathcal{F}_{h'}} (\|e\|_{\mathcal{H}_e} - \sqrt{6\omega})^2$

$Q_{\mathcal{T}_h} \geq 0 \dots$  parameter of “quality”

## Quality of mesh

- $Q_{\mathcal{T}_h} = \frac{1}{\#\mathcal{T}_h} \sum_{e \in \mathcal{T}_h} (\|e\|_{\mathcal{H}_e} - \sqrt{6\omega})^2 \geq 0$
- smaller  $Q_{\mathcal{T}_h}$  means “better” mesh.

## Idea of mesh optimization

modify locally mesh in such a way that  $Q_{\mathcal{T}_h}$  is decreasing

## Mesh optimization algorithm

- several local operations (adding a node, removing an edge, moving a node, etc.)
- tested and performed if  $Q_{\mathcal{T}_h}$  is decreasing

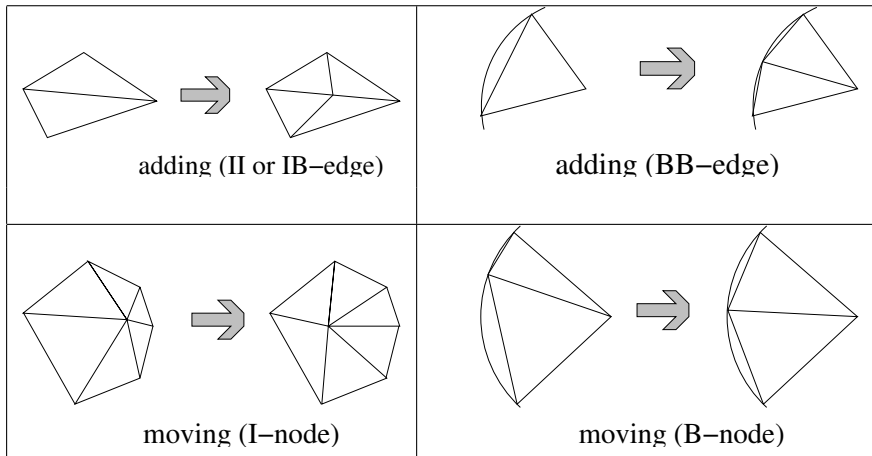
## Riemann metric

### Optimal mesh

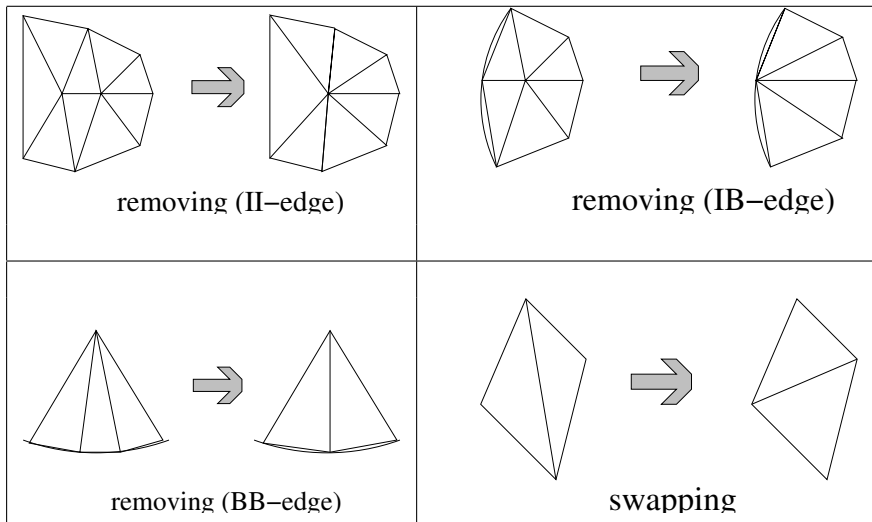
⇔ mesh is uniform in the Riemann metric generated by  $\mathcal{H}$



# Local operations: adding and moving



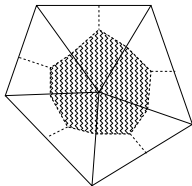
# Local operations: removing and swapping



## Mesh optimization algorithm

For given  $u \in V$  and  $\omega > 0$ , we can construct optimal mesh  $\mathcal{T}_h$

- $u$  has to be approximated by  $u_h$
- in practice, we need to approximate  $\mathcal{H}(u)$  on each edge only
- we approximate  $\mathcal{H}(u)$  at vertices  $P_k$  of mesh
- let  $D_k$  be a polygon around  $P_k$



$$\begin{aligned}\frac{\partial^2 u}{\partial x_i \partial x_j}(P_k) &\approx \frac{1}{|D_k|} \int_{D_k} \frac{\partial^2 u}{\partial x_i \partial x_j} dx = \frac{1}{|D_k|} \int_{\partial D_k} \frac{\partial u}{\partial x_i} n_j dS \\ &\approx \frac{1}{|D_k|} \int_{\partial D_k} \frac{\partial u_h}{\partial x_i} n_j dS\end{aligned}$$

## Approximation of $\mathcal{H}$

- previous approximation gives  $\mathcal{H}(u(P_k)) \approx \mathbb{H}(u_h(P_k))$
- $\mathbb{H}$  is symmetric, not positively definite
- we put  $\bar{\mathbb{H}} := |\mathbb{H}|$  using the eigenvalue decomposition

## Regularization

- if  $u_h$  is linear then  $\bar{\mathbb{H}} = 0 \dots$  problem
- if  $u_h$  is discontinuous then  $\bar{\mathbb{H}}$  can blow up
- in order to overcome this problem, we set

$$\mathbb{M}(P_k) = c \left[ \mathbb{I} + \frac{\varepsilon_1}{\varepsilon_1/p + \|\bar{\mathbb{H}}(P_k)\|} \bar{\mathbb{H}}(P_k) \right] \quad (2)$$

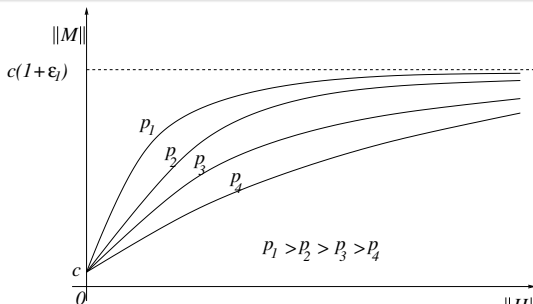
- we replace  $\bar{\mathbb{H}}$  by  $\mathbb{M}$
- $c, \varepsilon_1, p$  suitably chosen constants

# Setting of matrices $\mathbb{M}$ (metric)

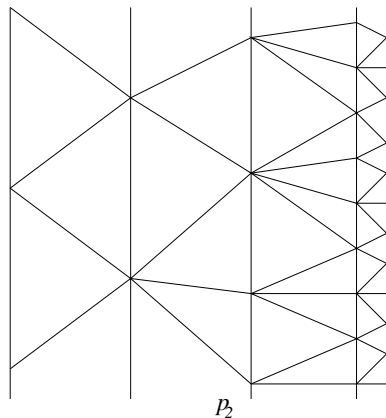
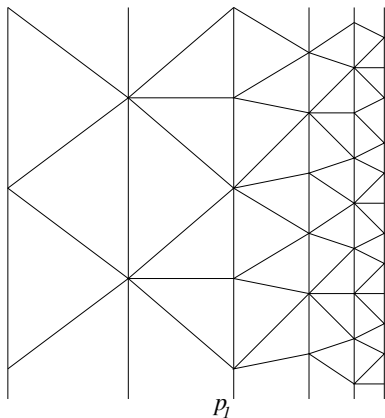
$$\mathbb{M}(P_k) = c \left[ \mathbb{I} + \frac{\varepsilon_1}{\varepsilon_1/p + \|\bar{\mathbb{H}}(P_k)\|} \bar{\mathbb{H}}(P_k) \right]$$

## Setting of constant

- if  $\bar{\mathbb{H}} = 0$  then  $\mathbb{M} = c\mathbb{I}$ ,  $c$  density of the coarsest mesh  $\sim \text{numel}$
- if  $\|\bar{\mathbb{H}}\| \rightarrow \infty$  then  $\|\mathbb{M}\| \rightarrow c(1 + \varepsilon_1) \approx c\varepsilon_1$ ,  $\varepsilon_1 = \left(\frac{\ell_{\max}}{\ell_{\min}}\right)^2$
- $p$  the “speed” of transition (from coarse to fine parts)



# Setting of matrices $\mathbb{M}$ (metric)



$$p_1 > p_2$$

# Multilevel computation

