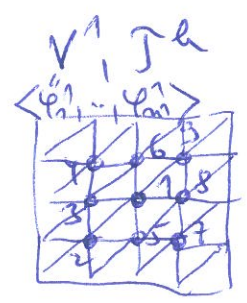
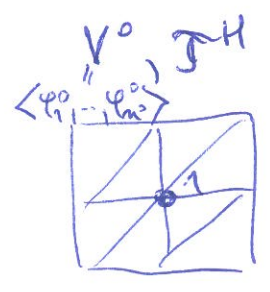


Metoda dvou sítí (two-grid)

(P) $\begin{cases} \text{Hledám } u(x) \in H_0^1(\Omega) \\ \int_{\Omega} k(x) \nabla u(x) \cdot \nabla v(x) dx = \int_{\Omega} f(x) v(x) dx \quad \forall v(x) \in H_0^1(\Omega) \end{cases}$

2 MKP diskrétnice



$V^0 \subset V^1$
 $A^0, A^1, \dots, p \times d$

$A^0 \bar{u}^0 = \bar{b}^0$
 \updownarrow
 $u^0(x) \in V^0 \quad b_0 \in (V^0)^*$

$A^1 \bar{u}^1 = \bar{b}^1$
 \updownarrow
 $u^1(x) \in V^1 \quad b_1 \in (V^1)^*$

Interpolace z hrubé sítě

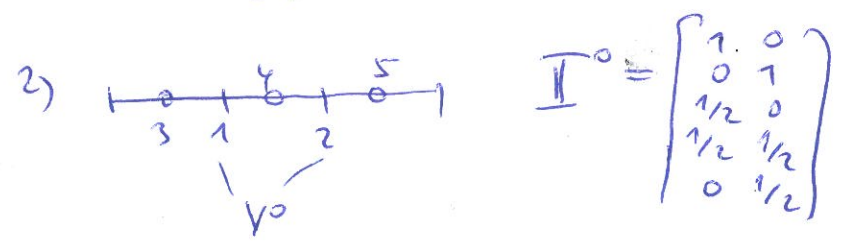
$I^0: V^0 \rightarrow V^1$ def. $(I^0(u^0(x)))(x) := u^0(x) \quad \forall u^0(x) \in V^0$

\forall MKP-souřadnice:

\bar{v}^0, \bar{v}^1
 $u^0(x) = \sum_i (\bar{v}^0)_i \varphi_i^0(x)$
 $u^1(x) = \sum_i (\bar{v}^1)_i \varphi_i^1(x)$

$(I^0)_{ij} = \varphi_j^0(x_i^1)$, $\forall \bar{v}^0 \mapsto I^0 \cdot \bar{v}^0$

např. 1) $I^0 = \begin{bmatrix} 1 \\ 1/2 \\ 1/2 \\ 1/2 \end{bmatrix}$ pro 2d příklad



Projekce rezidua (před → zpět) na hrubou síť

Lemma. $V^0 \xrightarrow{I^0} V^1 \Rightarrow (V^0)^* \xrightarrow{(I^0)^T} (V^1)^*$

Důk. $b^1 \in (V^1)^*$ je definováno přes V^1 a dluží si na $V^0 \in V^1$

$$\|b^1\|_{(V^0)^*} = \sup_{\substack{v^0 \in V^0 \\ \|v^0\|_V = 1}} \langle b^1, v^0 \rangle \leq \sup_{\substack{v^1 \in V^1 \\ \|v^1\|_V = 1}} \langle b^1, v^1 \rangle = \|b^1\|_{(V^1)^*}$$

MRP-soudadnice:

$$(\bar{b}^0)_i := \langle b^0, \varphi_i^0(x) \rangle = \langle b^1, I^0(\varphi_i^0(x)) \rangle = \bar{b}^1 \cdot (I^0 \cdot \bar{e}_i) = (I^0)^T \bar{b}^1$$

$$\bar{b}^1 \mapsto \boxed{\bar{b}^0 := (I^0)^T \bar{b}^1}$$

Korekce na hrubé síti (coarse-grid correction)

A-ortonormální projekce na hrubou síť:

$$u^1 \mapsto \underbrace{P^0 u^1}_{= \bar{u}^0} \stackrel{\text{def.}}{\Leftrightarrow} a(P^0 u^1, w^0) = a(u^1, w^0) \quad \forall w^0 \in V^0$$

$$\boxed{A^0 \bar{u}^0 = (I^0)^T A^1 u^1}$$

Chyba projekce (řád)

$$\bar{u}^1 - I^0 P^0 \bar{u}^1 = \boxed{[I - I^0 (A^0)^{-1} (I^0)^T A^1] \bar{u}^1}$$

Vyhlazení na jemné síti (smoother)

$\hat{A}^1 \approx A^1$ tak, že $\boxed{(\hat{A}^1)^{-1}}$ je klemé, např. $(\hat{A}^1)^{-1} = \frac{1}{\Lambda} I$

kde $\lambda_{\max}(A) \leq \Lambda$

Metoda dvou sítí (1 iterace ≈ předpokládáno)

$A^0 \bar{e}_0$

$\hat{u}_{1+m}^1 := T(\bar{r}_0^1):$

$\hat{u}_1^1 := I^0 (A^0)^{-1} (I^0)^T \bar{r}_0^1$ (korekce)

$\hat{r}_1^1 := \bar{r}_0^1 - A^1 \hat{u}_1^1 = A^1 [I - I^0 (A^0)^{-1} (I^0)^T A^1] \bar{e}_0$

$\hat{r}_1^1 \stackrel{A^1 \hat{e}_1}{=} \bar{r}_1^1$

for $i=1:m$

\hat{v}_{1+i-1}^1

$\hat{u}_{1+i}^1 := \hat{u}_{1+i-1}^1 + (\hat{A}^1)^{-1} \hat{r}_{1+i-1}^1$

$\hat{r}_{1+i}^1 := \hat{r}_{1+i-1}^1 - A^1 \hat{v}_{1+i-1}^1$

end for

Richardson + T.

$\bar{u}_0^1 = \bar{0}, \bar{r}_0^1 = \bar{b}^1$

$k := 0$

while $\|\bar{r}_k^1\| / \|\bar{b}^1\| > \epsilon$

$\bar{u}_{k+1}^1 := \bar{u}_k^1$

$+ T(\bar{r}_k^1)$

$\bar{r}_{k+1}^1 := \bar{r}_k^1 - A^1 \bar{u}_{k+1}^1$

end while

Metoda dvoju sliki (pohac.)

Veta.

• $\forall \bar{w}^1: \bar{w}^1 A^1 \bar{w}^1 \leq \bar{w}^1 \hat{A}^1 \bar{w}^1 \iff \|\bar{w}^1\|_A \leq \|\bar{w}^1\|_{\hat{A}}$
(smoothing property)

• $\exists K > 0 \forall \bar{w}^1:$
 $\|(I - P^0) \bar{w}^1\|_{\hat{A}} \leq K \|(I - P^0) \bar{w}^1\|_A$
(approximation property)

$$\| \bar{u}^1 - \hat{u}^1_{1+m} \|_A$$

$$\sqrt{K \gamma(m)} \cdot \| \bar{u}^1 - \hat{u}^1_0 \|_A$$

gdje

$$\gamma(m) = \frac{1}{2m+1} \left(1 - \frac{1}{2m+1} \right)^{2m}$$

Dokaz.

• Bi-ortogonalnost baze

$A := A^1, \hat{A} := \hat{A}^1 \dots \text{opd} \Rightarrow \hat{A}^{-1/2} A \hat{A}^{-1/2} \dots \text{opd}$

$$\hat{A}^{-1/2} A \hat{A}^{-1/2} \bar{v}_i = \lambda_i \bar{v}_i$$

$$=: \bar{\psi}_i$$

$0 \leq \lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$

$\bar{v}_1 \perp \bar{v}_2 \perp \dots \perp \bar{v}_n$ --ortonormirana baze \mathbb{R}^n

Pred: $\bar{v}_i \cdot \bar{v}_j = \delta_{ij} \implies (\hat{A}^{1/2} \bar{\psi}_i) \cdot (\hat{A}^{1/2} \bar{\psi}_j) = \bar{\psi}_i \hat{A} \bar{\psi}_j = \delta_{ij}$

$\bar{\psi}_i A \bar{\psi}_j = \lambda_i \delta_{ij}$

$(\bar{\psi}_i)$ je tzv. bi-ortogonalna baza vektora k A a \hat{A}

$$\|\bar{u}\|_A^2 = \left(\sum_i a_i \bar{\psi}_i \right) \cdot A \cdot \left(\sum_j a_j \bar{\psi}_j \right) = \sum_i \lambda_i (a_i)^2$$

$$\|\bar{u}\|_{\hat{A}}^2 = \left(\sum_i a_i \bar{\psi}_i \right) \hat{A} \left(\sum_j a_j \bar{\psi}_j \right) = \sum_i (a_i)^2$$

$$\hat{A}^{-1} A \bar{\psi}_i = \lambda_i \bar{\psi}_i$$

• Smoothing step

$$\bar{b} - A \hat{u}_{n+i-1} = A(\bar{u} - \hat{u}_{n+i-1})$$

$$\hat{u}_{n+i} = \hat{u}_{n+i-1} + (\hat{A})^{-1} \bar{b} - A \hat{u}_{n+i-1}$$

$$\hat{e}_{n+i} = \bar{u} - \hat{u}_{n+i} = \bar{u} - \hat{u}_{n+i-1} - (\hat{A})^{-1} A \hat{e}_{n+i-1} = (I - \hat{A}^{-1} A) \hat{e}_{n+i-1}$$

$$\hat{e}_{1+m} = (I - \hat{A}^{-1} A)^m \hat{e}_1$$

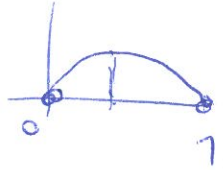
$$\|\hat{e}_{1+m}\|_A^2 = \left\| \sum_i (1 - \lambda_i)^m b_i \bar{\psi}_i \right\|_A^2 = \sum_i \lambda_i (1 - \lambda_i)^{2m} (b_i)^2$$

Smooth prop. $\forall \bar{v}: \|\bar{v}\|_A \leq \|\bar{v}\|_{\hat{A}} \Leftrightarrow \lambda_{\max}(\hat{A}^{-1}A) \leq 1$ (4)

$$\bar{v}^T A \bar{v} \leq \bar{v}^T \hat{A} \bar{v}$$

↳ $\lambda_i \in (0, 1)$

↳ $\max_i \lambda_i (1 - \lambda_i)^{2m} \leq \max_{\lambda \in (0, 1)} \lambda (1 - \lambda)^{2m} =: f(\lambda)$



$$f'(\lambda) = (1 - \lambda)^{2m} - 2m\lambda(1 - \lambda)^{2m-1} \stackrel{!}{=} 0$$

$$1 - \lambda = 2m\lambda \quad \lambda =$$

$$\lambda = \frac{1}{1 + 2m} : f(\lambda) = \frac{1}{1 + 2m} \left(1 - \frac{1}{1 + 2m}\right)^{2m} =: \gamma(m)$$

$$\|\hat{e}_{1+m}\|_A^2 \leq \gamma(m) \cdot \sum (b_i)^2 = \gamma(m) \cdot \|\hat{e}_1\|_A^2 \leq \gamma(m) \|(I - P^0)\bar{e}_0\|_A^2$$

$$\begin{aligned} &\leq K \gamma(m) \|(I - P^0)\bar{e}_0\|_A^2 \stackrel{\text{approx. } P \rightarrow P}{\leq} K \|(I - P^0)v\|_A^2 \leq K \|(I - P^0)v\|_A \\ &\stackrel{\text{Prop. 1}}{\leq} K \gamma(m) \|\bar{e}_0\|_A^2 \end{aligned}$$

□

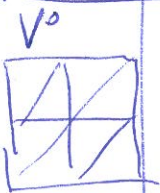
Smoothing property $\|\bar{v}\|_A \leq \|\bar{v}\|_{\hat{A}}$

režisérna verž. $\hat{A} = \frac{1}{\Lambda} I$, kde $\lambda_{\max}(A) \leq \Lambda$

Approximedian property $\|(I - P^0)\bar{v}\|_{\hat{A}} \leq K \|(I - P^0)\bar{v}\|_A$

plyne $P \rightarrow H^2$ -regularní úlohy z Aubin-Nitscheho leme. (příště)

Metoda více úrovní (multigrid)



$$V^L : A^L \bar{u}^L = \bar{b}^L$$

$$\bar{u}^l := M^l(\bar{r}^l)$$

if $l=0$

$$\bar{u}^0 := (A^0)^{-1} \bar{r}^0$$

else

$$\bar{u}_1^l = I^{l-1} \cdot M^{l-1}((I^{l-1})^T \bar{r}^l)$$

for $\bar{r}=1, p$ (same subc.)

$$\bar{u}_i^l := \bar{u}_i^{l-1} + M^{l-1}((I^{l-1})^T \bar{r}_{i-1}^l)$$

$$\bar{r}_i^l := \bar{r}_{i-1}^l - A^l \bar{u}_i^l$$

end for

for $i=1:m$ (smoothing)

$$\bar{u}_{pti}^l := \bar{u}_{pti-1}^l + (\hat{A}^l)^{-1} \bar{r}_{pti-1}^l$$

$$\bar{r}_{pti}^l := \bar{r}_{pti-1}^l - A^l \bar{v}_{pti-1}^l$$

Důsledek Aubin-Nitscheho lemma: approx. property

$\hat{A} := \Delta \cdot I$, kde $\Delta \geq \lambda_{\max}(A)$, ~~2d úloha~~ dimenze $d=2,3$

\Downarrow

~~$\mathbb{R}^k \rightarrow \mathbb{R}$~~

v_i
číslo podmíněnosti $M \otimes A$

$\|\bar{w}\|_{\hat{A}}^2 = \bar{w} \hat{A} \bar{w} = \Delta \cdot \|\bar{w}\|^2 \leq \frac{CA h^{d-2}}{c_1 h^d} \int_{\Omega} (w^a(x))^2$

$\bar{w} = \bar{u} - P\bar{u}$

\bar{u} ... MKP-souřadnice $u^1(x)$ ~~úloha~~ na $V^1 \subset V$

$P\bar{u}$... MKP-souřadnice $u^0(x) =$ Galerkinova projekce $u^1(x)$ na $I^0(V^0)$

Tedy $\|(I-P)\bar{u}\|_{\hat{A}}^2 \leq \frac{CA}{c_1 h^2} \int_{\Omega} (u^1(x) - u^0(x))^2 dx$ $\bar{u} A \bar{u}$

Důsledek A-N lemma

$|a(u,u)| \geq m \|u\|_1^2$

$\leq \frac{CA}{c_1 h^2} M^2 C^2 \underbrace{\|u^1 - u^0\|_1}_{H^2} \leq \underbrace{\frac{4CA^2 C^2}{c_1}}_{=:K} \|(I-P)\bar{u}\|_A^2$

Zaměňme-li v 2-grid algoritmu

pořadí korekce na hrubé sítki a zhlazování na jemné sítki,

pak platí lze ukázat, že platí:

$\bar{e}_{k+1} = (I-P)(I-\hat{A}^{-1}A)^m \bar{e}_k$

$\hat{q} := T(\bar{r}_0)$

$\| \bar{e}_{k+1} \|_A \leq \frac{K}{\sqrt{m}} \| \bar{e}_k \|_A$

$\bar{u}_0 := \bar{o}$
 for $i=1:m$
 $\bar{u}_i := \bar{u}_{i-1} + \hat{A}^{-1} \bar{r}_{i-1} =: \bar{v}_i$
 $\bar{r}_i := \bar{r}_{i-1} - A \bar{v}_i$
 end for
 $\hat{u}_{m+1} := \bar{u}_m + I^0(A^0)^{-1}(I^0)^T \bar{r}_m$
 $\hat{q} := \bar{u}_{m+1}$

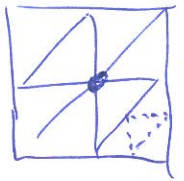
Metoda vice sídi (multigrid)

$$V^0 \subset V^1 \subset \dots \subset V^L$$

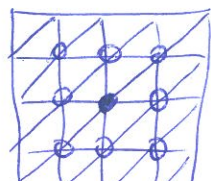
$$V^0: A^0$$

$$V^1: A^1, \hat{A}^1$$

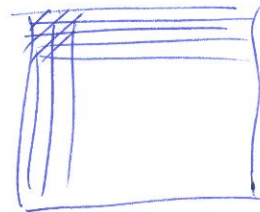
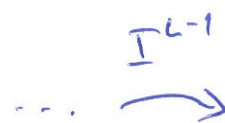
$$V^L: A^L, \hat{A}^L \hat{=} \hat{A}^1$$



$$A^0 \bar{u}^0 = \bar{b}^0$$



$$A^1 \bar{u}^1 = \bar{b}^1$$



$$A^L \bar{u}^L = \bar{b}^L$$

$\hat{=} \bar{u} \quad \hat{=} \bar{b}$

Multigrid: $\bar{v} := M_P^l(\bar{r}_0)$

if $l=0$

$$\bar{v} := (A_0)^{-1} \bar{r}_0$$

else

$$\bar{u}_0 := \bar{0}$$

for $i=1:m$

$$\bar{u}_i := \bar{u}_{i-1} + \underbrace{(\hat{A}^l)^{-1} \bar{r}_{i-1}}_{=: \bar{v}_i}$$

$$\bar{r}_i := \bar{r}_{i-1} - A^l \bar{v}_i$$

end for

for $i=1:p$

$$\bar{u}_{m+i} := \bar{u}_{m+i-1} + \underbrace{I^{l-1} M_P^{l-1} ((I^{l-1})^T \bar{r}_{m+i-1})}_{=: \bar{v}_i}$$

$$\bar{r}_{m+i} := \bar{r}_{m+i-1} - A^l \bar{v}_i$$

end for

$$\bar{v} := \bar{u}_{m+p}$$

Richardson

$$\bar{u}_{k+1} := \bar{u}_k + M_P^l(\bar{r}_k)$$

$$\bar{u}_{m+p} := \bar{u}_m + \bar{q}_p$$

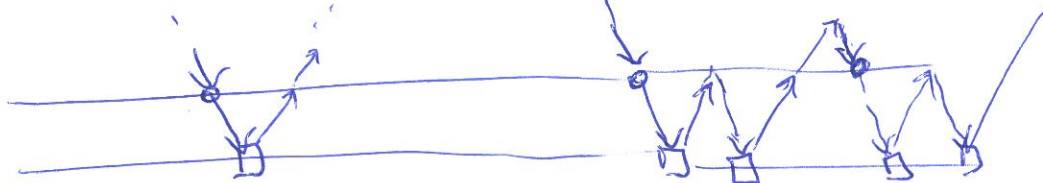
$p=1$: V-cykklus

$p=2$: W-cykklus

$l=L$
 $l=L-1$



$l=L-1$
 $l=0$



Plati:

$$\|\hat{q} - \bar{q}_1^l\|_A = \|\mathbf{I}^{l-1} \{ \mathbf{M}_P^{l-1} - (\mathbf{A}^{l-1})^{-1} \} (\mathbf{I}^{l-1})^T \bar{\mathbf{r}}_m^l\|_A$$

$$= \|(\mathbf{M}_P^{l-1} \mathbf{A}^{l-1} - \mathbf{I}) \bar{\mathbf{r}}_m^{l-1}\|_A$$

$$\|\hat{q} - \bar{q}_P^l\|_A = \|(\mathbf{M}_P^{l-1} \mathbf{A}^{l-1} - \mathbf{I})^P \bar{\mathbf{r}}_m^{l-1}\|_A$$

Věta. (konvergence multigridu)

$$\left. \begin{array}{l} \bullet p \geq 2 \\ \bullet \gamma \in (0, 1) \end{array} \right\} \Rightarrow \forall m \geq \left\lceil \frac{K}{\gamma - \gamma^p} \right\rceil^2 : \|\bar{\mathbf{r}}_{k+1}^l\| \leq \gamma \|\bar{\mathbf{r}}_k^l\|_A$$

nesdílí na l.

Důkaz. (indukcí)

$$l=0: \bar{\mathbf{r}}_{k+1}^l = \bar{\mathbf{0}}$$

$$l>0: \bar{\mathbf{r}}_{k+1}^l = \bar{\mathbf{u}}^l - \bar{\mathbf{u}}_{m+p}^l = \bar{\mathbf{u}}^l - \bar{\mathbf{u}}_m^l - \bar{\mathbf{q}}_P^l \pm \hat{\mathbf{q}}^l$$

$$\|\bar{\mathbf{r}}_{k+1}^l\|_A \leq \underbrace{\|\bar{\mathbf{u}}^l - \bar{\mathbf{u}}_m^l - \hat{\mathbf{q}}^l\|_A}_{\leq \frac{K}{\sqrt{m}} \|\bar{\mathbf{r}}_k^l\|_A} + \underbrace{\|\hat{\mathbf{q}}^l - \bar{\mathbf{q}}_P^l\|_A}_{\leq \gamma^p \|\bar{\mathbf{r}}_k^l\|_A} \leq \gamma \|\bar{\mathbf{r}}_k^l\|_A$$

$$\leq \frac{K}{\sqrt{m}} \|\bar{\mathbf{r}}_k^l\|_A$$

$$\leq \gamma^p \|\bar{\mathbf{r}}_k^l\|_A$$

konvergence
2-gridu

indukční předpoklad
+ předchozí pozorování

Výpočetní náročnost 1 iterace multigridu

$$W^l \leq C \cdot m \cdot m^k + p \cdot W^{l-1} \leq C m m^k + p [C m m^{k-1} + p W^{l-2}] \leq \dots$$

$$\leq C m [m^k + p m^{k-1} + \dots + p^{l-1} m^1 + K]$$

Předp.
 $m^l \geq \underbrace{D}_{\geq 2} \cdot m^{l-1}$

$$CPU((A^0)^{-1})$$

$$\leq C m m^k \left[1 + \left(\frac{p}{D}\right) + \left(\frac{p}{D}\right)^2 + \dots + \left(\frac{p}{D}\right)^{l-1} \right] + C m K$$

$$\leq \sum_{i=0}^{l-1} q^i \in \mathbb{R} \quad p \text{ w } q := \left\lfloor \frac{p}{D} < 1 \right\rfloor$$

A tedy $CPU(M_P^k(\dots)) = O(m^k)$