

# Computational comparison of the discretization and iteration errors

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# Iteration error

▶  $-\Delta u = f$  in  $\Omega$ ,  $u = 0$  on  $\partial\Omega$

▶  $u_h \in V_h$ : 
$$\underbrace{\int_{\Omega} \nabla u_h \cdot \nabla v_h \, dx}_{a(u_h, v_h)} = \underbrace{\int_{\Omega} f v_h \, dx}_{F(v_h)} \quad \forall v_h \in V_h$$

▶  $u_h = \sum_{i=1}^{N_{DOF}} y_i \varphi_i \quad Ay = b \quad A_{ij} = a(\varphi_j, \varphi_i)$   
 $b_i = F(\varphi_i)$

▶  $u_h^* = \sum_{i=1}^{N_{DOF}} y_i^* \varphi_i \quad Ay^* = b^*$

▶  $\|u - u_h^*\|_a^2 = \|u - u_h\|_a^2 + 2a(u - u_h, u_h - u_h^*) + \|u_h - u_h^*\|_a^2$

discretization error:  $\|u - u_h\|_a \approx O(h)$

iteration error:  $\|u_h - u_h^*\|_a \approx O(?)$

$$\|u\|_a^2 = a(u, u)$$

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▶  $\|u - u_h^*\|_a^2 = \|u - u_h\|_a^2 + \|u_h - u_h^*\|_a^2$

discretization error:  $\|u - u_h\|_a \approx O(h)$

iteration error:  $\|u_h - u_h^*\|_a \approx O(?)$

$$\|u\|_a^2 = a(u, u)$$

# Upper bound for the iteration error



## Lemma

If  $Ay = b$ ,  $Ay^* = b^*$ , and  $\kappa(A) = \|A\| \|A^{-1}\|$  then

$$\frac{\|y - y^*\|}{\|y\|} \leq \kappa(A) \frac{\|b - Ay^*\|}{\|b\|}.$$

## Proof.

$$\begin{aligned} \frac{\frac{\|y - y^*\|}{\|y\|}}{\frac{\|b - Ay^*\|}{\|b\|}} &= \frac{\frac{\|A^{-1}(b - b^*)\|}{\|y\|}}{\frac{\|b - b^*\|}{\|Ay\|}} = \frac{\|A^{-1}(b - b^*)\|}{\|b - b^*\|} \frac{\|Ay\|}{\|y\|} \\ &\leq \|A^{-1}\| \|A\| \equiv \kappa(A) \end{aligned}$$



# Choice of norm



$$\frac{\|u_h - u_h^*\|_a}{\|u_h\|_a} = \frac{\|y - y^*\|_A}{\|y\|_A} \leq \kappa_A(A) \frac{\|b - Ay^*\|_A}{\|b\|_A} \quad \|y\|_A^2 = y^T Ay$$

$$\frac{\|u_h - u_h^*\|_{L^2(\Omega)}}{\|u_h\|_{L^2(\Omega)}} = \frac{\|y - y^*\|_M}{\|y\|_M} \leq \kappa_M(A) \frac{\|b - Ay^*\|_M}{\|b\|_M} \quad \|y\|_M^2 = y^T My$$

$$\frac{\|u_h - u_h^*\|_{L^\infty(\Omega)}}{\|u_h\|_{L^\infty(\Omega)}} = \frac{\|y - y^*\|_\infty}{\|y\|_\infty} \leq \kappa_\infty(A) \frac{\|b - Ay^*\|_\infty}{\|b\|_\infty} \quad \|y\|_\infty = \max_i |y_i|$$

$$? = \frac{\|y - y^*\|_{\ell^2}}{\|y\|_{\ell^2}} \leq \kappa_{\ell^2}(A) \frac{\|b - Ay^*\|_{\ell^2}}{\|b\|_{\ell^2}} \quad \|y\|_{\ell^2}^2 = y^T y$$

$$\|u_h\|_a = \|y\|_A \quad \|u_h\|_{L^2(\Omega)} = \|y\|_M \quad \|u_h\|_{L^\infty(\Omega)} = \|y\|_\infty \quad ? = \|y\|_{\ell^2}$$

# Condition number in $M$ -norm



## Lemma

$$A, M \text{ s.p.d} \Rightarrow \kappa_M(A) = \|A^{-1}\|_M \|A\|_M \leq \kappa_{\ell^2}(M) \kappa_{\ell^2}(A)$$

## Proof.

$$\blacktriangleright M \text{ s.p.d.} \Rightarrow \|M^{-1}\| = \sup_{0 \neq y \in \mathbb{R}^N} \frac{y^T M^{-1} y}{y^T y} = \sup_{\substack{0 \neq z \in \mathbb{R}^N \\ z = M^{-1/2} y}} \frac{z^T z}{z^T M z}$$

$$\begin{aligned} \blacktriangleright \|A\|_M^2 &= \sup_{0 \neq z \in \mathbb{R}^N} \frac{\|Az\|_M^2}{\|z\|_M^2} = \sup_{0 \neq z \in \mathbb{R}^N} \frac{z^T A M A z}{z^T M z} \\ &= \sup_{0 \neq z \in \mathbb{R}^N} \frac{z^T A M A z}{z^T A^2 z} \frac{z^T A^2 z}{z^T z} \frac{z^T z}{z^T M z} \leq \|M\| \|A^2\| \|M^{-1}\| \end{aligned}$$

$$\blacktriangleright \|A^{-1}\|_M^2 \leq \|M\| \|A^{-2}\| \|M^{-1}\|$$

$$\blacktriangleright \kappa_M^2(A) \leq \kappa^2(M) \kappa(A^2) = \kappa^2(M) \kappa^2(A) \quad \|\cdot\| = \|\cdot\|_{\ell^2}$$



# Numerical example – continuous problem



$$\begin{aligned} -\Delta u &= f & \text{in } \Omega &= (0, 1)^2 \\ u &= 0 & \text{on } \partial\Omega \end{aligned}$$

▶  $-\Delta \tilde{v}_{kl} = \tilde{\lambda}_{kl} \tilde{v}_{kl}$

▶  $\tilde{v}_{kl} = 2 \sin(k\pi x_1) \sin(\ell\pi x_2)$

$$\int_{\Omega} v_{kl} v_{mn} = \delta_{(kl)(mn)}$$

$$\tilde{\lambda}_{kl} = \pi^2(k^2 + \ell^2)$$

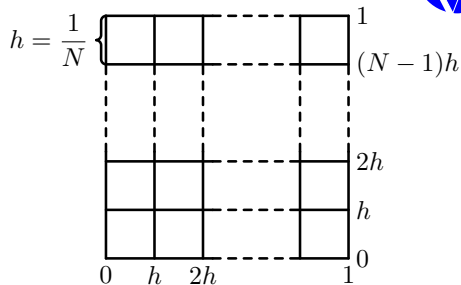
$$k, \ell = 1, 2, \dots$$

▶  $f = \sum_{k,\ell} c_{kl} \tilde{\lambda}_{kl} \tilde{v}_{kl} \quad \Rightarrow \quad u = \sum_{k,\ell} c_{kl} \tilde{v}_{kl}$

# Numerical example – discretization



$$\blacktriangleright A = \frac{1}{3} \times \begin{array}{cccc} & -1 & -1 & -1 \\ & | & | & | \\ -1 & -1 & 8 & -1 \\ & | & | & | \\ & -1 & -1 & -1 \end{array}$$



$$\blacktriangleright Aw^{k\ell} = \lambda_{k\ell} w^{k\ell} \quad k, \ell = 1, 2, \dots, N-1$$

$$\blacktriangleright w_{ij}^{k\ell} = h \tilde{v}(ih, jh) = 2h \sin(k\pi ih) \sin(\ell\pi jh) \quad i, j = 1, 2, \dots, N-1$$

$$\lambda_{k\ell} = \frac{2}{3} \left( 4 - \cos(k\pi h) - \cos(\ell\pi h) - 2 \cos(k\pi h) \cos(\ell\pi h) \right)$$

$$w^{k\ell} \cdot w^{mn} = \delta_{(k\ell)(mn)}$$

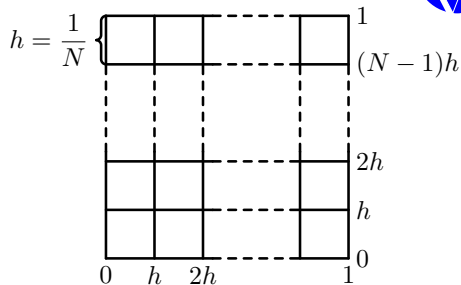
$$\blacktriangleright \kappa(A) = \frac{2 + \cos^2(\pi h)}{2 - \cos(\pi h) - \cos^2(\pi h)} \approx O(h^{-2})$$



# Numerical example – discretization



$$\blacktriangleright M = \frac{h^2}{36} \times \begin{array}{c} 1 \text{ --- } 4 \text{ --- } 1 \\ | \qquad | \qquad | \\ 4 \text{ --- } 16 \text{ --- } 4 \\ | \qquad | \qquad | \\ 1 \text{ --- } 4 \text{ --- } 1 \end{array}$$



$$\blacktriangleright Mw^{k\ell} = \mu_{k\ell} w^{k\ell} \qquad k, \ell = 1, 2, \dots, N-1$$

$$\blacktriangleright w_{ij}^{k\ell} = \text{dtto.} \qquad i, j = 1, 2, \dots, N-1$$

$$\blacktriangleright \mu_{k\ell} = \frac{h^2}{9} \left( 4 + 2 \cos(k\pi h) + 2 \cos(\ell\pi h) + \cos(k\pi h) \cos(\ell\pi h) \right)$$

$$\blacktriangleright \kappa(M) = \frac{4 + 4 \cos(\pi h) + \cos^2(\pi h)}{4 - 4 \cos(\pi h) + \cos^2(\pi h)} \xrightarrow{h \rightarrow 0} 9$$

# Numerical example – discretization



- ▶  $\int_{\Omega} \tilde{\lambda}_{kl} \tilde{v}_{kl}(x_1, x_2) \varphi_{ij}(x_1, x_2) dx_1 dx_2 = d_{kl} w_{ij}^{kl}$
- ▶  $d_{kl} = \frac{k^2 + \ell^2}{k^2 \ell^2 \pi^2 h^3} 4 \left(1 - \cos(k\pi h)\right) \left(1 - \cos(\ell\pi h)\right)$
- ▶  $f = \sum_{k,l} c_{kl} \tilde{\lambda}_{kl} \tilde{v}_{kl} \Rightarrow u = \sum_{k,l} c_{kl} \tilde{v}_{kl}$   
 $b = \sum_{k,l} c_{kl} d_{kl} w^{kl} \Rightarrow y = \sum_{k,l} \frac{c_{kl} d_{kl}}{\lambda_{kl}} w^{kl}$   
 $\Rightarrow u_h = \sum_{i=1}^{N-1} \sum_{j=1}^{N-1} y_{ij} \varphi_{ij}$

$$c_{11} = 10 \quad c_{23} = 2 \quad c_{33} = 1$$

$$\blacktriangleright \|u\|_a^2 = \sum_{k,l} c_{kl}^2 \tilde{\lambda}_{kl} \qquad \|u\|_a \approx O(1)$$

$$\blacktriangleright \|u_h\|_a^2 = \|y\|_A^2 = \sum_{k,l} c_{kl}^2 d_{kl}^2 / \lambda_{kl} \qquad \|u_h\|_a \approx O(1)$$

$$\blacktriangleright \|u - u_h\|_a^2 = \|u\|_a^2 - \|u_h\|_a^2 \qquad \|u - u_h\|_a \approx O(h)$$

$$\blacktriangleright \|f\|_a^2 = \sum_{k,l} c_{kl}^2 \tilde{\lambda}_{kl}^3 \qquad \|f\|_a \approx O(1)$$

$$\blacktriangleright \|b\|_A^2 = \sum_{k,l} c_{kl}^2 d_{kl}^2 \lambda_{kl} \qquad \|b\|_A \approx O(h^2)$$

$$d_{kl} \approx O(h) \quad \lambda_{kl} \approx O(h^2)$$

# $L^2$ norms



$$\triangleright \|u\|_{L^2(\Omega)}^2 = \sum_{k,l} c_{kl}^2 \qquad \|u\|_{L^2(\Omega)} \approx O(1)$$

$$\triangleright \|u_h\|_{L^2(\Omega)}^2 = \|y\|_M^2 = \sum_{k,l} \frac{c_{kl}^2 d_{kl}^2 \mu_{kl}}{\lambda_{kl}^2} \qquad \|u_h\|_{L^2(\Omega)} \approx O(1)$$

$$\triangleright \int_{\Omega} uu_h \, dx_1 \, dx_2 = \sum_{k,l} \frac{c_{kl}^2 d_{kl}^2}{\lambda_{kl} \tilde{\lambda}_{kl}} \qquad \approx O(1)$$

$$\triangleright \|u - u_h\|_{L^2(\Omega)}^2 = \|u\|_{L^2(\Omega)}^2 - 2 \int_{\Omega} uu_h \, dx_1 \, dx_2 + \|u_h\|_{L^2(\Omega)}^2$$
$$\|u - u_h\|_{L^2(\Omega)} \approx O(h^2)$$

$$\triangleright \|f\|_{L^2(\Omega)}^2 = \sum_{k,l} c_{kl}^2 \tilde{\lambda}_{kl}^2 \qquad \|f\|_{L^2(\Omega)} \approx O(1)$$

$$\triangleright \|b\|_M^2 = b^T M b = \sum_{k,l} c_{kl}^2 d_{kl}^2 \mu_{kl} \qquad \|b\|_M \approx O(h^2)$$

$$\triangleright \|b\|_{\ell^2}^2 = b^T b = \sum_{k,l} c_{kl}^2 d_{kl}^2 \qquad \|b\|_{\ell^2} \approx O(h)$$

# Conjugate gradient algorithm

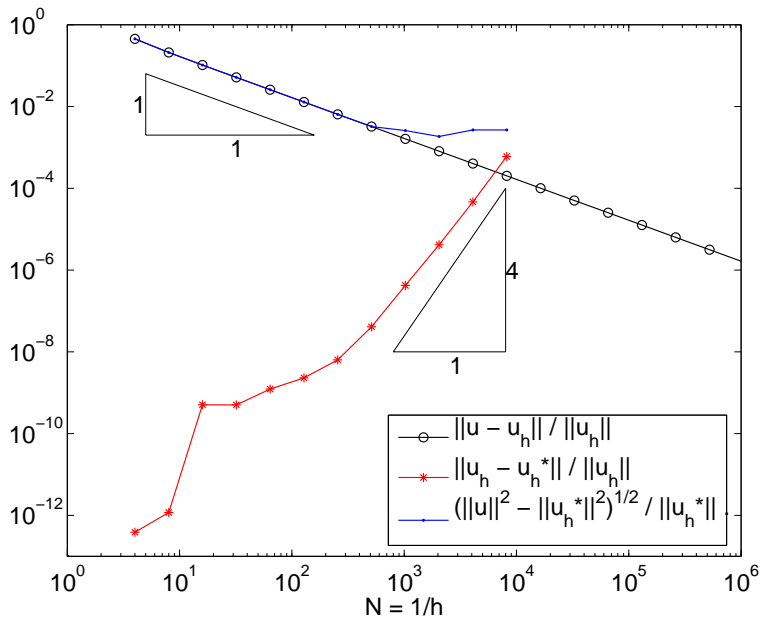


$$p_0 = r_0 = b - Ay_0$$

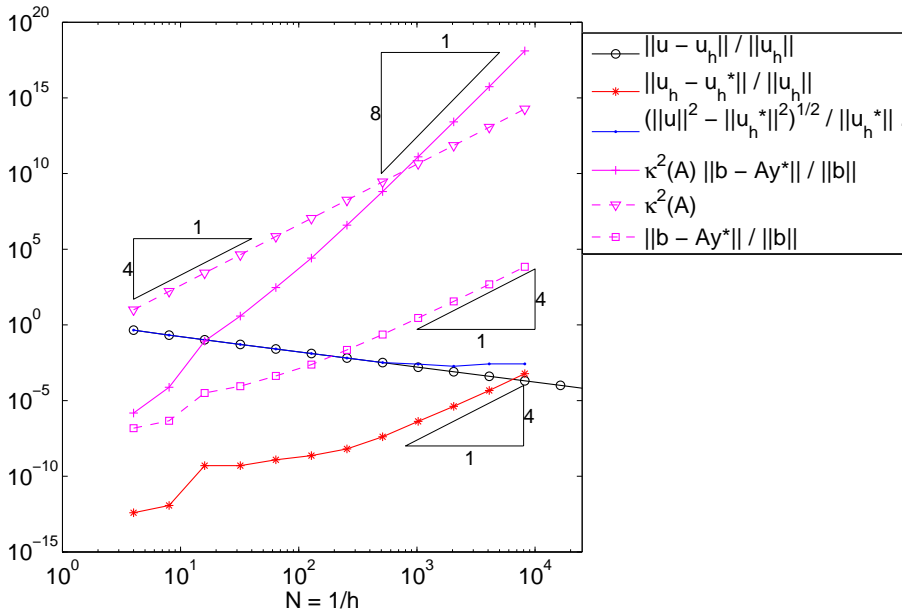
for  $k = 0, 1, 2, \dots$  do

- ▶ if  $\frac{r_k^T r_k}{b^T b} \leq TOL^2$  then STOP  $TOL = 10^{-4}$
- ▶  $\alpha_k = \frac{r_k^T r_k}{p_k^T A p_k}$  (step length)
- ▶  $y_{k+1} = y_k + \alpha_k p_k$  (approximate solution)
- ▶  $r_{k+1} = r_k - \alpha_k A p_k$  (residual)
- ▶  $\beta_k = \frac{r_{k+1}^T r_{k+1}}{r_k^T r_k}$  (residual improvement)
- ▶  $p_{k+1} = r_{k+1} - \beta_k p_k$  (search direction)

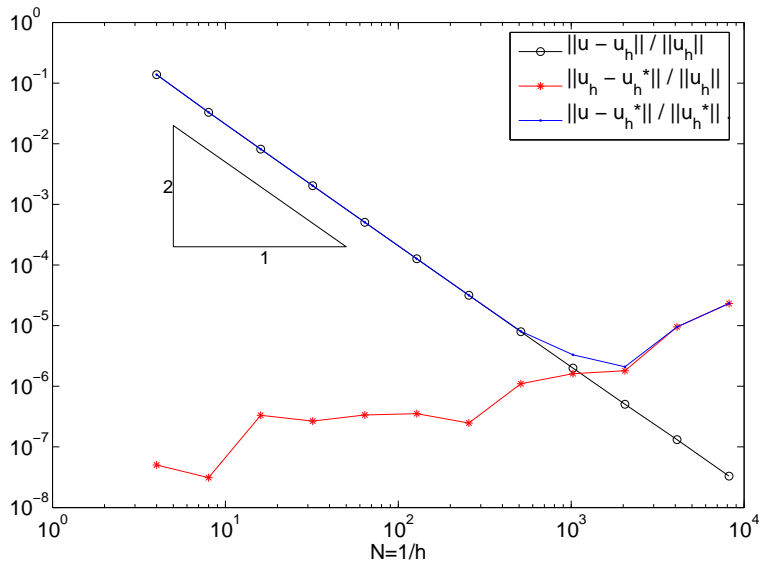
# Relative errors – energy norm – $\|\cdot\|_a$



# Relative err. and upper bound – energy norm – $\|\cdot\|_a$

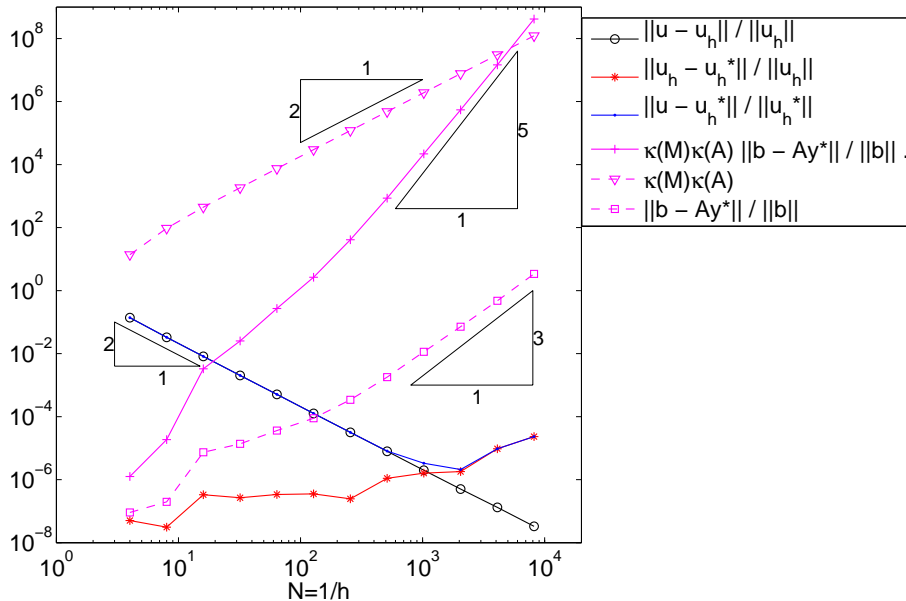


# Relative errors – $L^2(\Omega)$ norm – $\|\cdot\|_{L^2(\Omega)}$

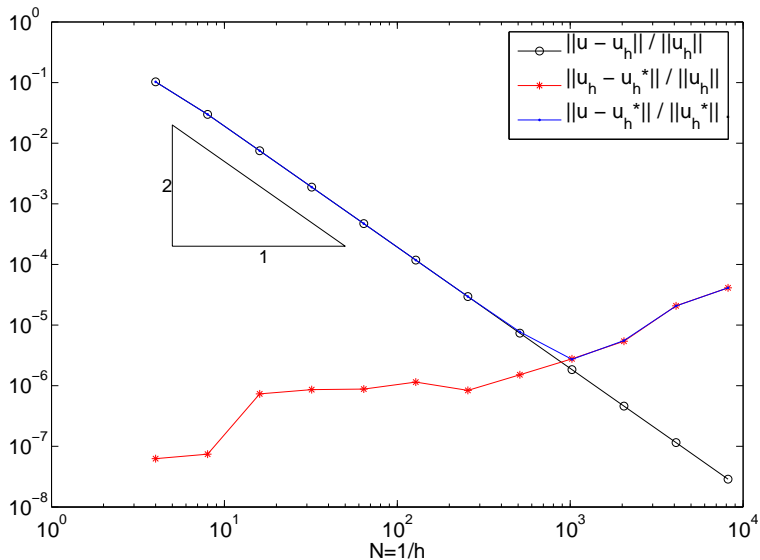




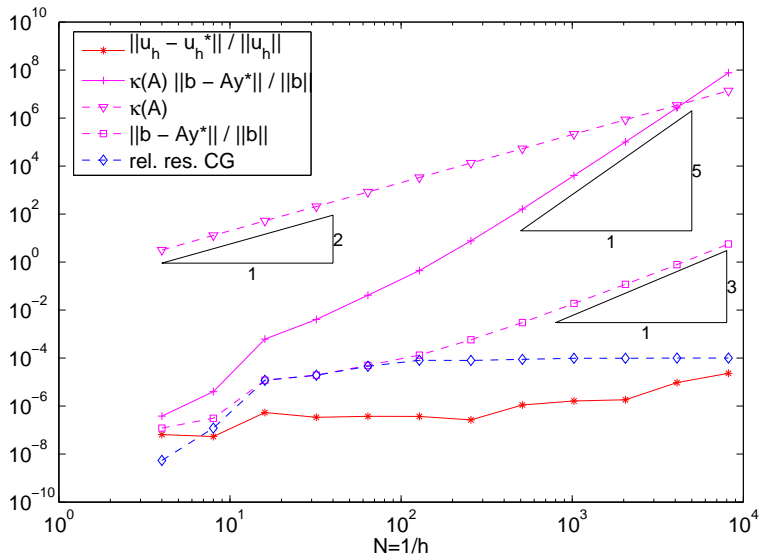
# Relative err. and upper bound – $L^2(\Omega)$ norm – $\|\cdot\|_{L^2(\Omega)}$



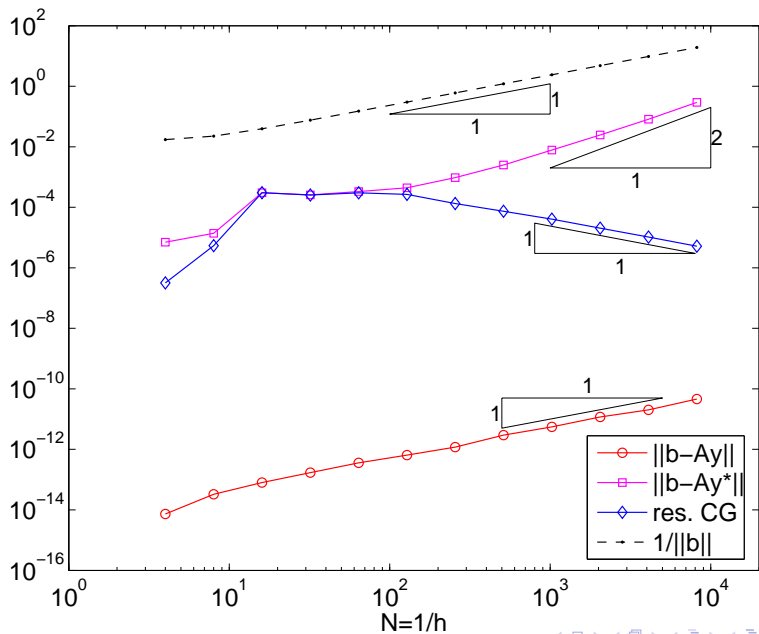
# Relative errors – discrete $\ell^\infty$ norm – $\|\cdot\|_\infty$



# Relative iteration error – Euclidean norm – $\|\cdot\|_{\ell^2}$



# Absolute residuals – Euclidean norm – $\|\cdot\|_{\ell^2}$



- ▶ iteration error – **single precision** – relevant from  $10^{6-8}$  DOFs

- ▶  $\frac{\|y - y^*\|}{\|y\|} \leq \kappa(A) \frac{\|b - Ay^*\|}{\|b\|}$     overestimation by  $10^{2-22}$

- ▶ Correct norm?
- ▶ Better stopping criterion for CG?
- ▶ Theory?

# Thank you for your attention

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