

On Estimating the A -norm of the Error in CG and PCG

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Consider a system of linear algebraic equations

$$Ax = b$$

where $A \in \mathbb{R}^{n \times n}$ is symmetric positive definite, $b \in \mathbb{R}^n$.

given $x_0 \in \mathbb{R}^n$, $r_0 = b - Ax_0$

Conjugate Gradient Method (CG)

$$x_j \in x_0 + \mathcal{K}_j(A, r_0)$$

$$\|x - x_j\|_A = \min_{u \in x_0 + \mathcal{K}_j(A, r_0)} \|x - u\|_A,$$

where

$$\mathcal{K}_j(A, r_0) \equiv \text{span } \{r_0, \dots, A^{j-1}r_0\}$$

$$\|x - x_j\|_A \equiv ((x - x_j), A(x - x_j))^{\frac{1}{2}}$$

Conjugate Gradient Method (CG) Hestenes and Stiefel (1952)

given $x_0, r_0 = b - Ax_0, p_0 = r_0,$

for $j = 0, 1, 2, \dots$

$$\gamma_j = \frac{(r_j, r_j)}{(p_j, Ap_j)}$$

$$x_{j+1} = x_j + \gamma_j p_j$$

$$r_{j+1} = r_j - \gamma_j Ap_j$$

$$\delta_{j+1} = \frac{(r_{j+1}, r_{j+1})}{(r_j, r_j)}$$

$$p_{j+1} = r_{j+1} + \delta_{j+1} p_j.$$

How to measure quality of approximation?

. . . it depends on what problem we solve.

- **using residual information,**

- normwise backward error,
- relative residual norm.

Hestenes and Stiefel: “Using of the residual vector r_j as a measure of the ‘goodness’ of the estimate x_j is not reliable” (HS 1952 p. 410)

- **using error estimates,**

- estimate of the A -norm of the error,
- estimate of the Euclidean norm of the error.

Hestenes and Stiefel: “The function $(x-x_j, A(x-x_j))$ can be used as a measure of the ‘goodness’ of x_j as an estimate of x .” (HS 1952 p. 413)

On Estimating the A -norm of the Error

Outline

1. Basic identity
2. Construction of estimate in CG
3. Estimates in finite precision arithmetic
4. Estimate of the A-norm of the error in PCG
5. Numerical experiments
6. Conclusions

1. Basic identity

At any iteration step j , CG (implicitly) determines weights and nodes of the j -point Gauss quadrature

$$\int_{\zeta}^{\xi} f(\lambda) d\omega(\lambda) = \sum_{i=1}^j \omega_i^{(j)} f(\theta_i^{(j)}) + R_j(f).$$

The Gauss quadrature for $f(\lambda) \equiv \lambda^{-1}$ gives

$$C_n = C_j + \frac{\|x - x_j\|_A^2}{\|r_0\|^2}, \quad C_n = \frac{\|x - x_0\|_A^2}{\|r_0\|^2}.$$

$C_n, C_j \dots$ continued fractions corresponding to $\omega(\lambda)$ and $\omega^{(j)}(\lambda)$
Golub: DGN-1978, GF-1993, GM-1994, GS-1994, GM-1997, ...

Equivalent formulas (multiplied by $\|r_0\|^2$)

In general (ST-2002):

$$\|x - x_0\|_A^2 = \|r_0\|^2 \text{Gauss } Q(j) + \|x - x_j\|_A^2.$$

W-2000:

$$r_0^T(x - x_0) = r_0^T(x_j - x_0) + \|x - x_j\|_A^2.$$

HS-1952, ST-2002:

$$\|x - x_0\|_A^2 = \sum_{i=0}^{j-1} \gamma_i \|r_i\|^2 + \|x - x_j\|_A^2.$$

This formula derived *purely algebraically* is equivalent to the Gauss quadrature formula!

2. Construction of estimate

Idea: Consider, for example,

$$\|x - x_j\|_A^2 = \|r_0\|^2 [C_n - C_j].$$

Run d extra steps. Subtracting identity for $\|x - x_{j+d}\|_A^2$ gives

$$\|x - x_j\|_A^2 = \|r_0\|^2 [C_{j+d} - C_j] + \|x - x_{j+d}\|_A^2.$$

When $\|x - x_j\|_A^2 \gg \|x - x_{j+d}\|_A^2$, we have a tight (lower) estimate.

GS-1994 and GM-1997

Mathematically equivalent estimates:

$$\|x - x_j\|_A^2 = EST^2 + \|x - x_{j+d}\|_A^2$$

GS-1994 and GM-1997

$$\eta_{j,d} = \|r_0\|^2 [C_{j+d} - C_j]$$

W-2000

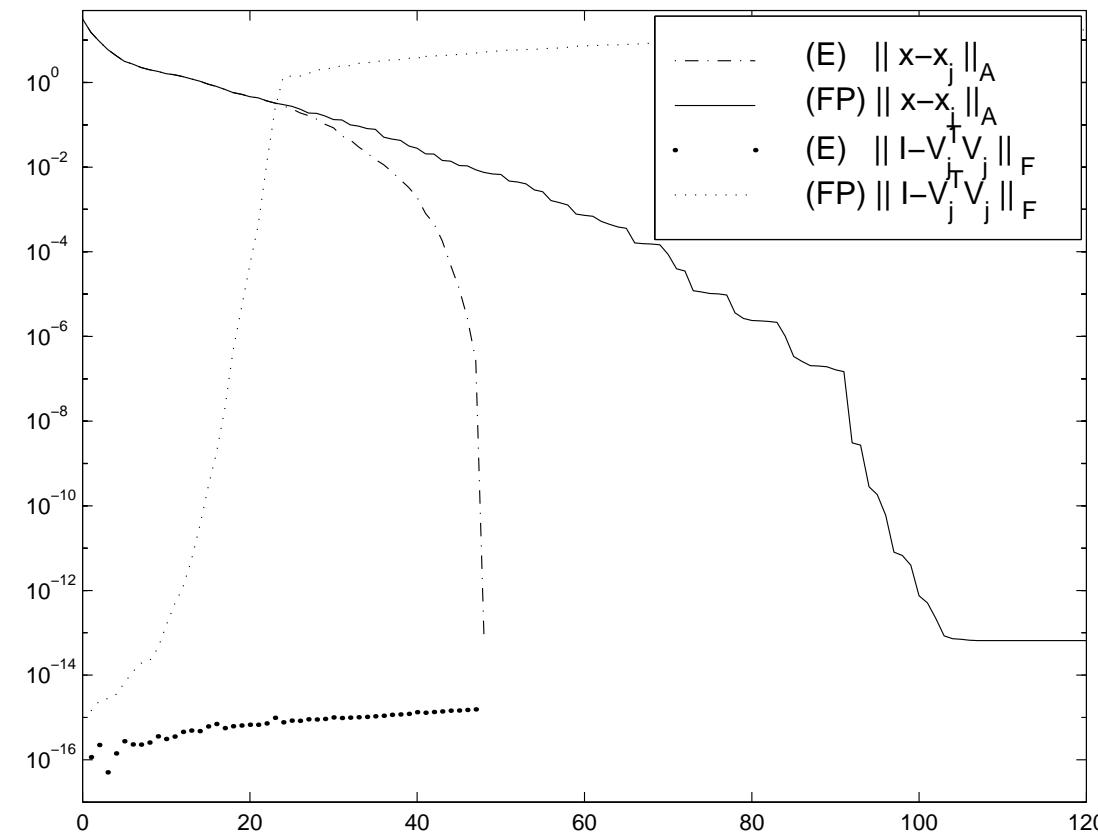
$$\mu_{j,d} = r_0^T (x_{j+d} - x_j)$$

HS-1952

$$\nu_{j,d} = \sum_{i=j}^{j+d-1} \gamma_i \|r_i\|^2$$

3. Estimates in finite precision arithmetic

orthogonality is lost, convergence is delayed!



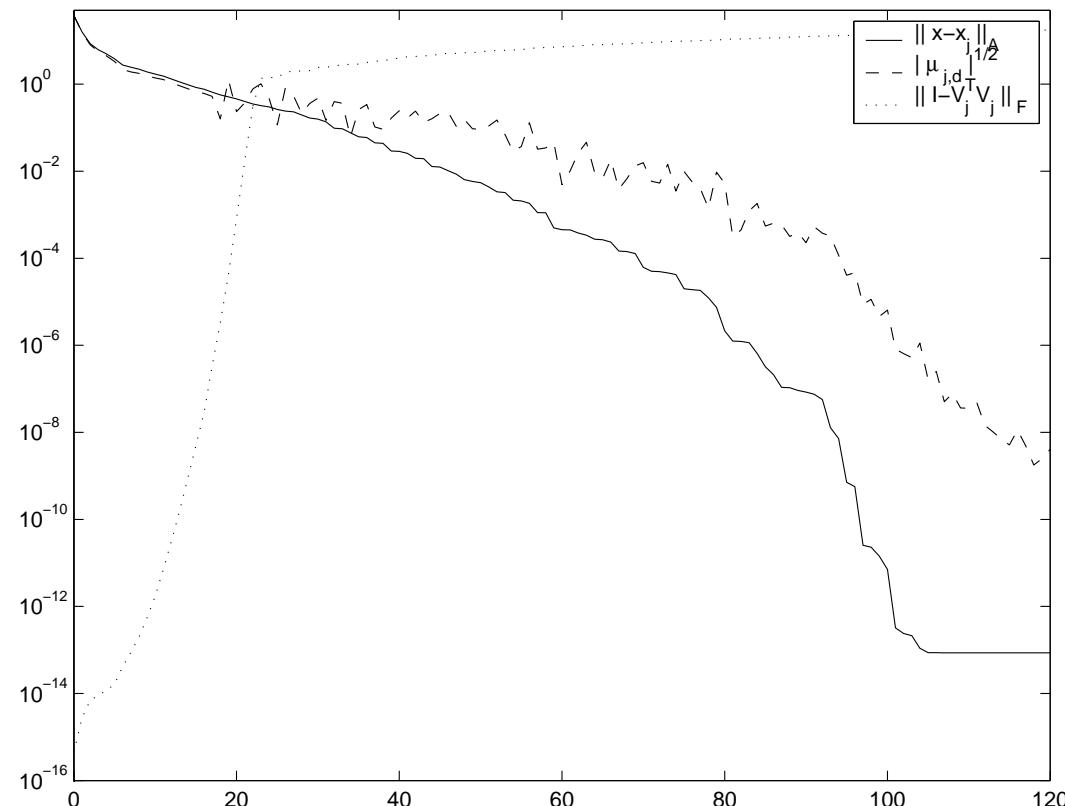
- The problem consists in validity of the whole identity

$$\|x - x_j\|_A^2 = EST^2 + \|x - x_{j+d}\|_A^2$$

An example:

$$\mu_{j,d} = r_0^T(x_{j+d} - x_j)$$

does not work!



Do the estimates give good information in practical computations?

$\eta_{j,d}$ **yes** GS-1994, GM-1997 (alg. **CGQL**)

Based on Greenbaum (1989), Greenbaum, S (1992), $\sqrt{\varepsilon}$ **limit**.

$\mu_{j,d}$ **no** ST-2002

$\nu_{j,d}$ **yes** ST-2002

Based on detailed analysis of preserving the local orthogonality in the HS algorithm, with using Paige (1971-80), Greenbaum (1989, 97)

4. Estimate in PCG

The cg-iterates are thought of being applied to

$$\hat{A}\hat{x} = \hat{b}$$

giving CG for $\hat{A}\hat{x} = \hat{b}$. We consider symmetric preconditioning

$$\hat{A} = L^{-1}AL^{-T}, \quad \hat{b} = L^{-1}b.$$

Definition

$$M \equiv LL^T, \quad \gamma_j \equiv \hat{\gamma}_j, \quad \delta_j \equiv \hat{\delta}_j,$$

$$x_j \equiv L^{-T}\hat{x}_j, \quad r_j \equiv L\hat{r}_j, \quad p_j \equiv L^{-T}\hat{p}_j, \quad s_j \equiv M^{-1}r_j :$$

Preconditioned Conjugate Gradient Method (PCG)

given $x_0, r_0 = b - Ax_0, s_0 = M^{-1}r_0, p_0 = s_0,$

for $j = 0, 1, 2, \dots$

$$\gamma_j = \frac{(r_j, s_j)}{(p_j, Ap_j)}$$

$$x_{j+1} = x_j + \gamma_j p_j$$

$$r_{j+1} = r_j - \gamma_j Ap_j$$

$$s_{j+1} = M^{-1}r_{j+1}$$

$$\delta_{j+1} = \frac{(r_{j+1}, s_{j+1})}{(r_j, s_j)}$$

$$p_{j+1} = s_{j+1} + \delta_{j+1} p_j.$$

How to estimate the \mathbf{A} -norm of the error in PCG?

Hestenes-Stiefel estimate for $\hat{A}\hat{x} = \hat{b}$:

$$\|\hat{x} - \hat{x}_j\|_{\hat{A}}^2 = \sum_{i=j}^{j+d-1} \hat{\gamma}_i \|\hat{r}_i\|^2 + \|\hat{x} - \hat{x}_{j+d}\|_{\hat{A}}^2.$$

Substitution $\hat{A} = L^{-1}AL^{-T}$, $\hat{x}_j = L^T x_j$, $\hat{\gamma}_i = \gamma_i$, $\hat{r}_i = L^{-1} r_i$ gives

$$\|\hat{x} - \hat{x}_j\|_{\hat{A}}^2 = (L^T x - L^T x_j) L^{-1} A L^{-T} (L^T x - L^T x_j) = \|x - x_j\|_A^2$$

and

$$\|x - x_j\|_A^2 = \sum_{i=j}^{j+d-1} \gamma_i (r_i, s_i) + \|x - x_{j+d}\|_A^2.$$

Numerical stability of the estimate in PCG.

ST-2003: rounding error analysis.

Analysis based on:

- rounding error analysis from ST-2002,
- solving of

$$Ms_{j+1} = r_{j+1}$$

enjoys perfect normwise backward stability (Hi-1996).

result:

Until $\|x - x_j\|_A$ reaches a level close to $\varepsilon \|x - x_0\|_A$,
the estimate $\nu_{j,d}$ **must work.**

Estimate of the relative A -norm of the error $\frac{\|x-x_j\|_A}{\|x-x_0\|_A}$.

Since

$$\begin{aligned}\|x - x_j\|_A^2 &= \nu_{j,d} + \|x - x_{j+d}\|_A^2, \\ \|x - x_0\|_A^2 &= \nu_{0,j+d} + \|x - x_{j+d}\|_A^2,\end{aligned}$$

the square root of

$$\varrho_{j,d} \equiv \frac{\nu_{j,d}}{\nu_{0,j+d}} = \frac{\|x - x_j\|_A^2 - \|x - x_{j+d}\|_A^2}{\|x - x_0\|_A^2 - \|x - x_{j+d}\|_A^2} \leq \frac{\|x - x_j\|_A^2}{\|x - x_0\|_A^2}$$

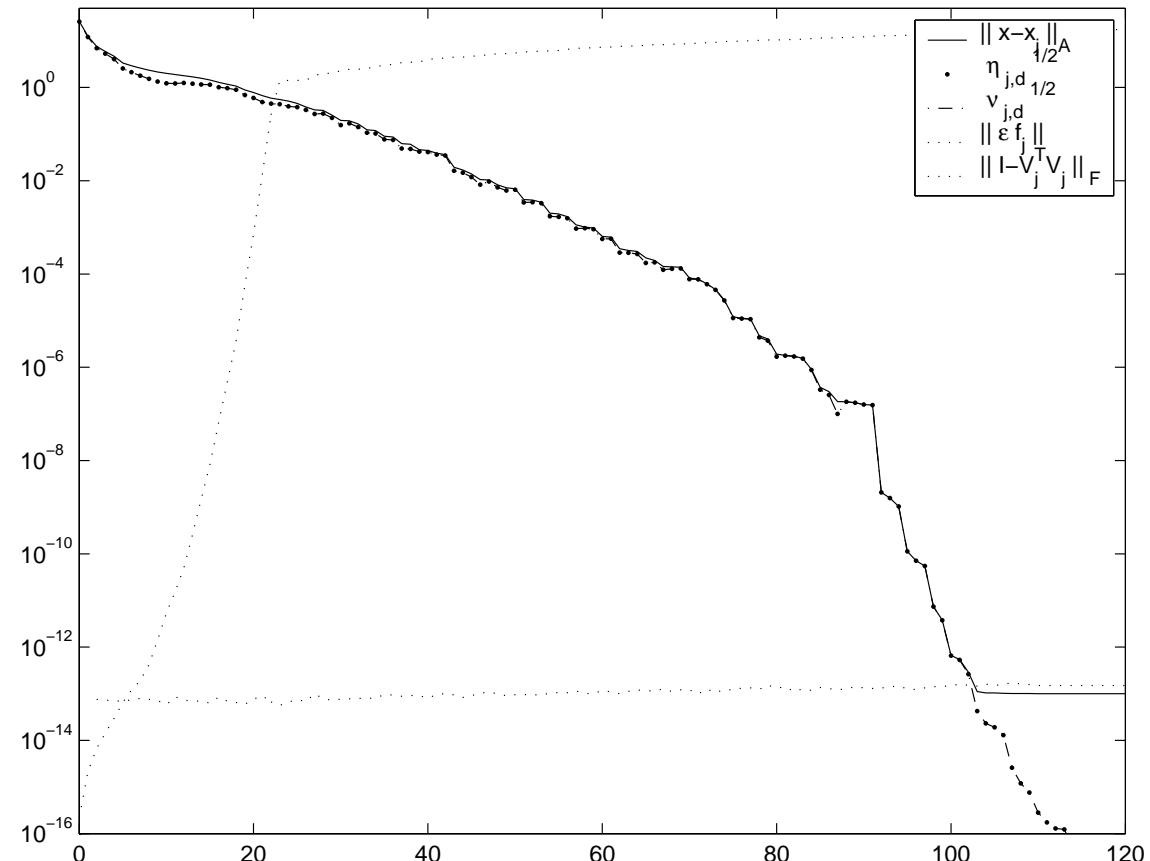
is a lower bound for the j th relative A -norm of the error.

5. Numerical Experiments

CG

Example from
Strakoš-1991

$n = 48$, $d = 4$,
 $\kappa(A) = 10000$
 $\lambda_{max} = 1000$

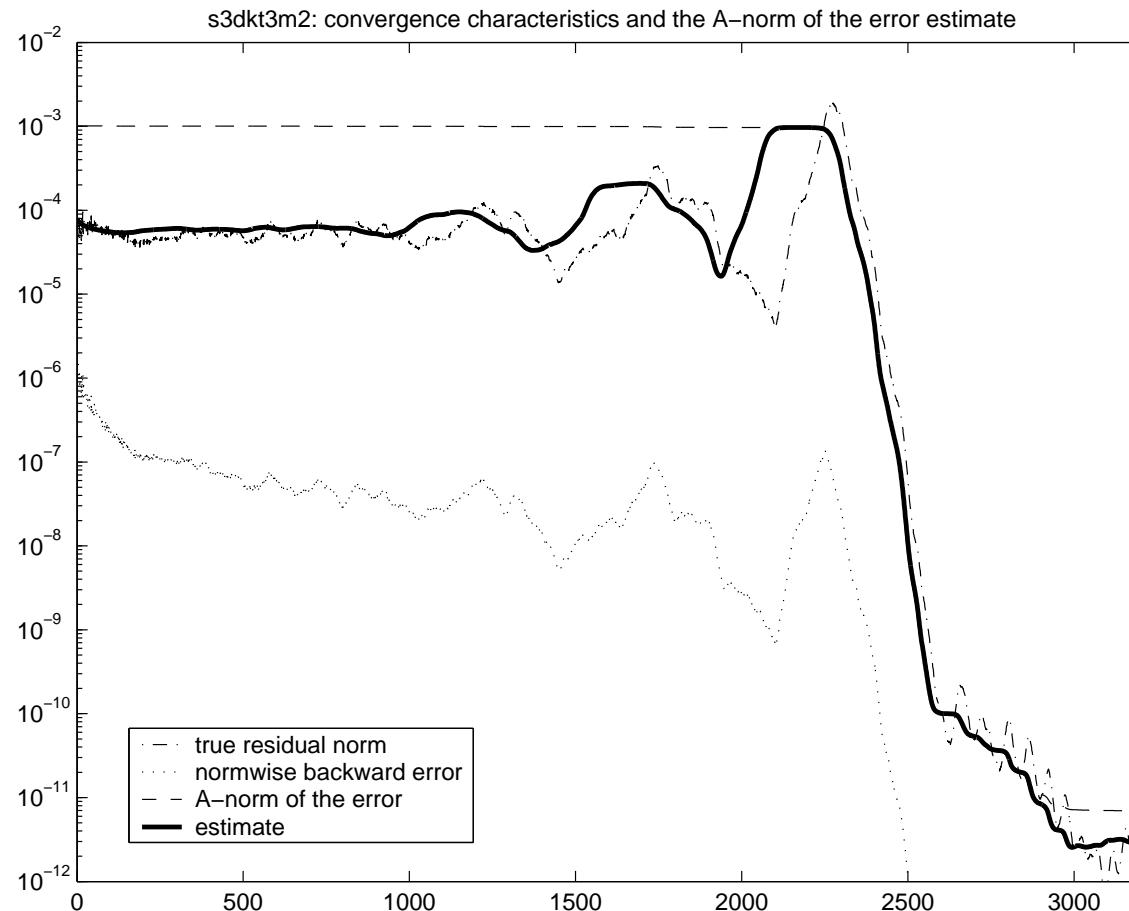


PCG

Matrix Market
Cylshell
matrix S3DKT3M2

$$\begin{aligned}\kappa(A) &= 3.62e + 11 \\ n &= 90499, d = 200 \\ L &= \text{cholinc}(A, 0)\end{aligned}$$

R. Kouhia: Cylindrical shell



PCG

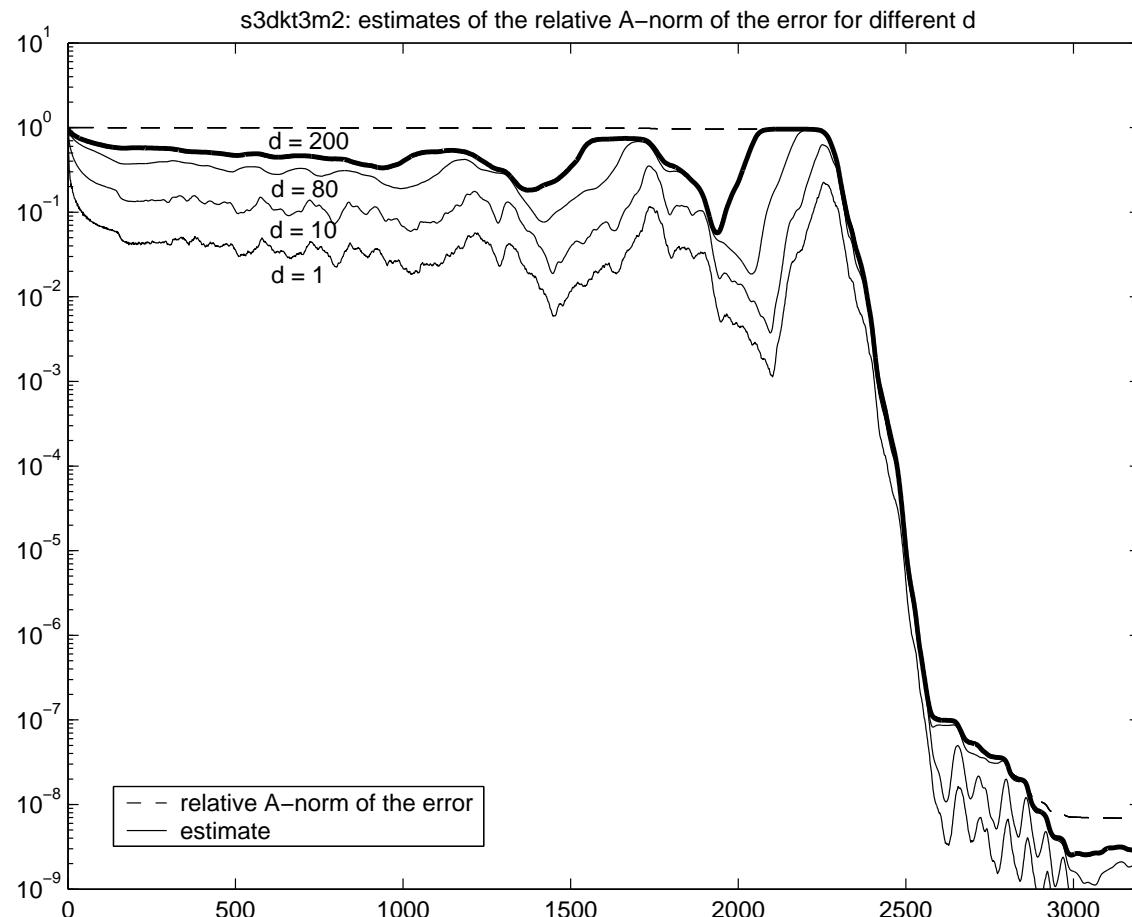
source:
Matrix Market
Collection Cylshell
matrix S3DKT3M2

$$\kappa(A) = 3.62e + 11$$

$$n = 90499,$$

$$L = \text{cholinc}(A, 0)$$

R. Kouhia: Cylindrical shell



PCG

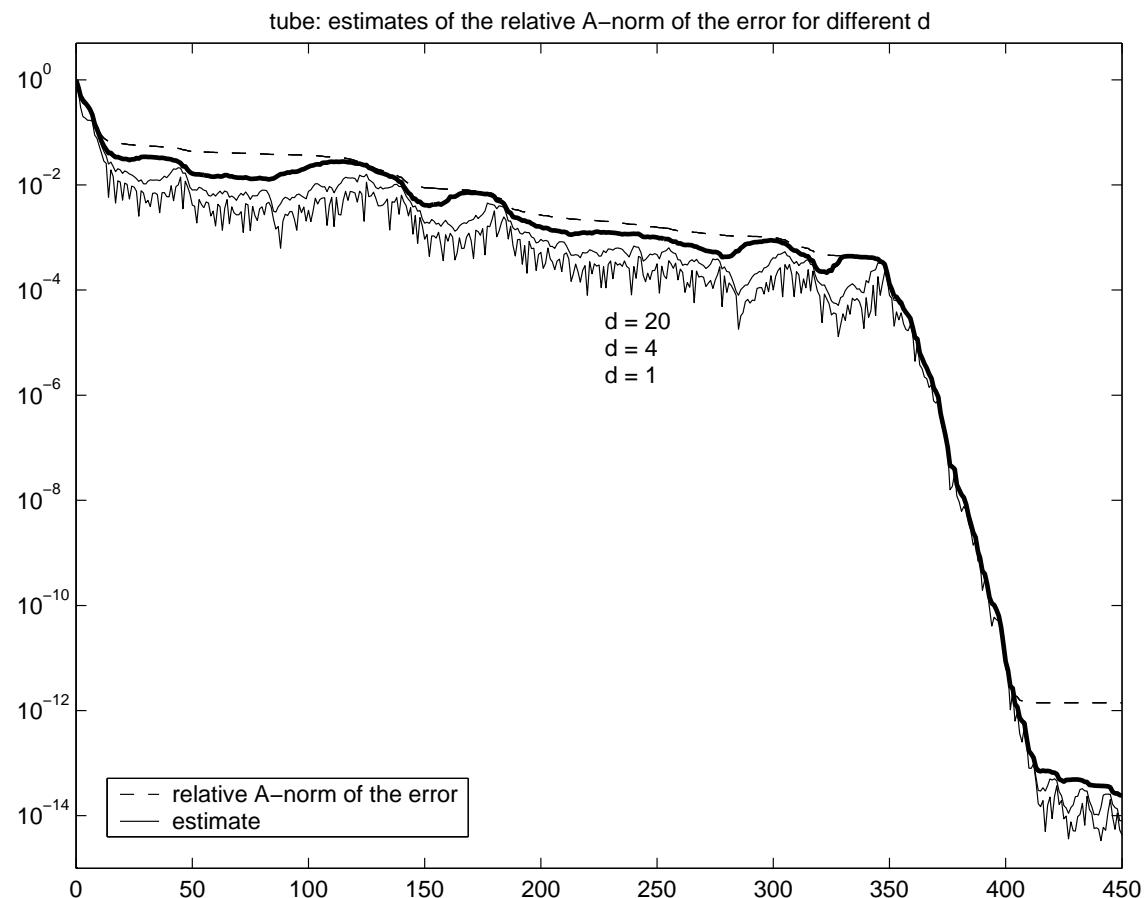
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R. Kouhia homepage
matrix tube1-2

$n = 21498$,

$L = \text{cholinc}(A, 1e-5)$

R. Kouhia: Cylindrical shell



6. Conclusions

- Up-to-now used bounds (based on Gauss quadrature) are **mathematically equivalent** to the formulas given (but somehow hidden) in the original Hestenes and Stiefel paper.
- Hestenes and Stiefel estimate is very simple, it can be computed almost for free and it is **numerically stable**.
- We suggest the estimates $\nu_{j,d}^{1/2}$ and $\varrho_{j,d}^{1/2}$ to be incorporated into **software realizations** of the PCG method.
- The estimates are tight if the A -norm of the error **reasonably decreases**.

Open problem: The adaptive choice of the parameter d .

More details can be found at

<http://www.cs.cas.cz/~strakos>

<http://www.cs.cas.cz/~tichy>

papers

- Strakoš, Z. and Tichý, P., *On Error Estimation in the Conjugate Gradient Method and Why It Works In Finite Precision Computations*, Electron. Trans. Numer. Anal. (ETNA), Volume 13, pp. 56-80, 2002.
- Strakoš, Z. and Tichý, P., *Simple estimation of the A-norm of the error in the Preconditioned Conjugate Gradient Method*, submitted to Comput. Methods Appl. Mech. Engrg., 2003.

technical report

- Strakoš, Z. and Tichý, P., *Simple estimation of the A-norm of the error in the Preconditioned Conjugate Gradient Method*, Technical Report No. 892, ICS AS CR, Prague, Czech Republic, 2003.