

# Behaviour of the Gauss-Radau upper bound

Petr Tichý

Charles University, Prague

joint work with

Gérard Meurant

GAMM Workshop, September 22, 2022, Prague

# The conjugate gradient method

$A$  is symmetric and positive definite,  $Ax = b$

**input**  $A, b$

$$r_0 = b, p_0 = r_0$$

**for**  $k = 1, 2, \dots$  until conv. **do**

$$\gamma_{k-1} = \frac{r_{k-1}^T r_{k-1}}{p_{k-1}^T A p_{k-1}}$$

$$x_k = x_{k-1} + \gamma_{k-1} p_{k-1}$$

$$r_k = r_{k-1} - \gamma_{k-1} A p_{k-1}$$

$$\delta_k = \frac{r_k^T r_k}{r_{k-1}^T r_{k-1}}$$

$$p_k = r_k + \delta_k p_{k-1}$$

**end for**

**Vectors**  $\in \mathcal{K}_k(A, b)$

$$\text{span}\{b, Ab, \dots, A^{k-1}b\}$$

Orthogonality

$$r_i \perp r_j \quad p_i \perp_A p_j$$

Coefficients  $\rightarrow T_k$

# How to measure quality of approximation?

... it depends on what problem we solve.

- **using residual information,**

- normwise backward error,
- relative residual norm.

[Hestenes, Stiefel 1952]: “Using of the residual vector  $r_k$  as a measure of the “goodness” of the estimate  $x_k$  is not reliable”

- **using error estimates,**

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

[Hestenes, Stiefel 1952] : “The function  $(x - x_k, A(x - x_k))$  can be used as a measure of the “goodness” of  $x_k$  as an estimate of  $x$ .”

# Estimating the $A$ -norm of the error in CG

$$\|x - x_k\|_A^2$$

- An important role in **stopping criteria**:

[Deuflhard 1994], [Arioli 2004],  
[Jiránek, Strakoš, Vohralík 2006], [Papež, Vohralík 2022]

- **Estimating errors** using **quadrature** approach:

[Dahlquist, Golub, Nash 1978],  
[Golub, Meurant 1994, 1997], [Golub, Strakoš 1994],  
[Meurant 1997, 1999, 2005], [Calvetti, Morigi, Reichel, Sgallari, 2000, 2001],  
[Strakoš, T. 2002], [Meurant, T. 2013, 2019], [Meurant, Papež, T. 2021]

- Why it works in **finite precision** arithmetic?

[Paige 1976, 1980, Greenbaum 1989],  
[Golub, Strakoš 1994], [Strakoš, T. 2002, 2005, 2011]

## Quadrature bounds

# Gauss quadrature lower bound

- It holds that

$$\gamma_k \|r_k\|^2 < \|x - x_k\|_A^2 \equiv \varepsilon_k.$$

- One can improve the accuracy of the lower bound using

$$\varepsilon_k = \underbrace{\sum_{j=k}^{\ell-1} \gamma_j \|r_j\|^2}_{\Delta_{k:\ell-1}} + \varepsilon_\ell.$$

[Golub, Strakoš 1994, Golub, Meurant 1997, Strakoš, T. 2002, 2005]

- How to choose  $\ell > k$  such that

$$\frac{\varepsilon_k - \Delta_{k:\ell-1}}{\varepsilon_k} \leq \tau.$$

[Meurant, Papež, T. 2021]

# Gauss-Radau upper bound

- Given  $\mu \leq \lambda_{\min}$ , it holds that

$$\|x - x_k\|_A^2 < \gamma_k^{(\mu)} \|r_k\|^2$$

where

$$\gamma_{k+1}^{(\mu)} = \frac{\left( \gamma_k^{(\mu)} - \gamma_k \right)}{\mu \left( \gamma_k^{(\mu)} - \gamma_k \right) + \delta_{k+1}}, \quad \gamma_0^{(\mu)} = \frac{1}{\mu}.$$

[Meurant, T. 2013]

## Practically relevant questions:

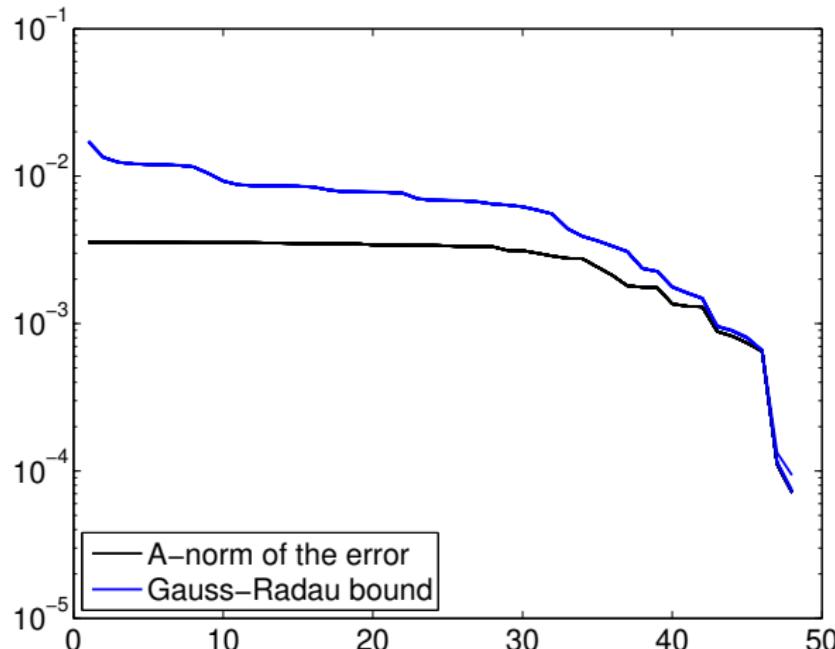
- How to get  $\mu$ ? [Gergelits, Mardal, Nielsen, Strakoš 2019, 2020, 2022]  
[Ladecký, Pultarová, Zeman 2021, 2021]
- Quality** of the bound?
- Numerical **behavior**?

## Behaviour of the upper bound

# Upper bound in exact arithmetic

Gauss-Radau bound, bcsstk01 matrix,  $n = 48$

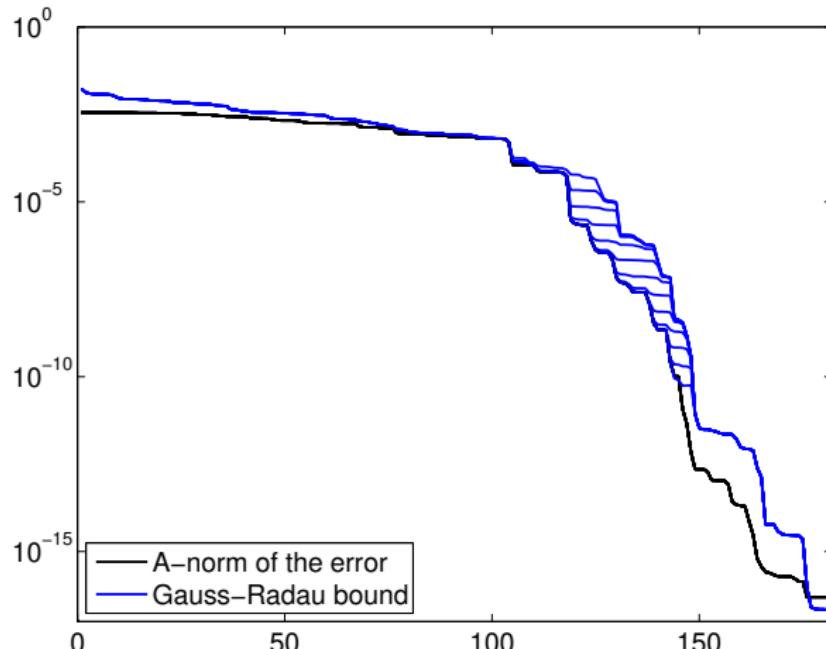
$$\mu = \frac{\lambda_{\min}}{1 + 10^{-m}}, \quad m = 2, \dots, 14$$



# Upper bound in finite precision arithmetic

Gauss-Radau bound, bcsstk01 matrix,  $n = 48$

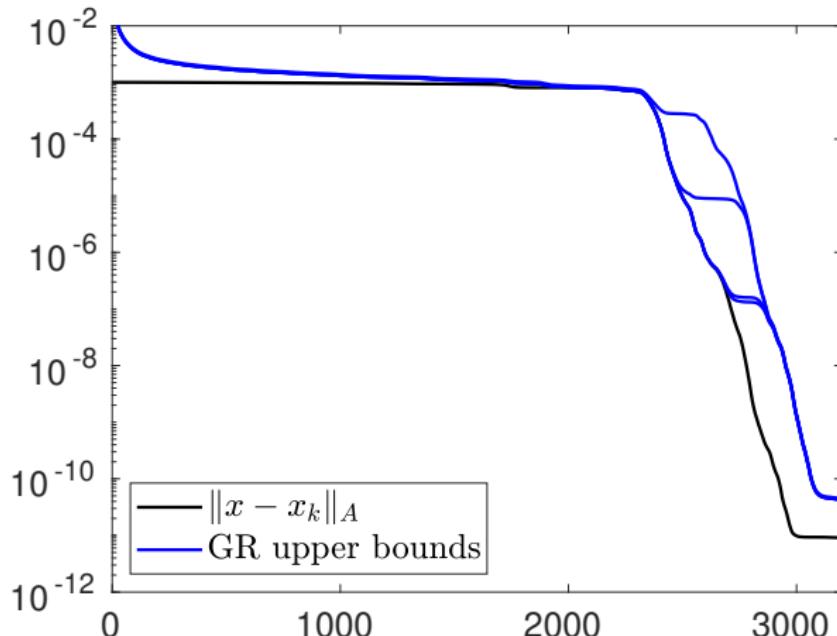
$$\mu = \frac{\lambda_{\min}}{1 + 10^{-m}}, \quad m = 2, \dots, 14$$



# Upper bound in finite precision arithmetic

Gauss-Radau bound, s3dkt3m2 matrix,  $n = 90449$

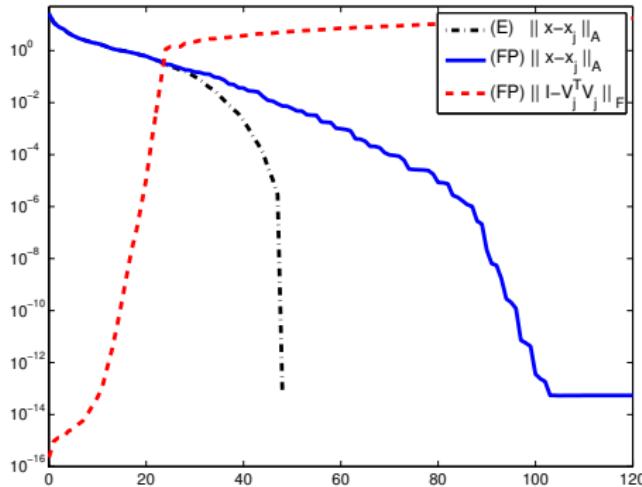
$$\gamma_{k+1}^{(\mu)} = \frac{(\gamma_k^{(\mu)} - \gamma_k)}{\mu (\gamma_k^{(\mu)} - \gamma_k) + \delta_{k+1}}$$



# Mathematical model of finite precision CG computations

The results of **finite precision CG** can be interpreted (up to a small inaccuracy) as the results of **exact CG** applied to a larger problem with a matrix having **clustered eigenvalues** around  $\lambda_i$ 's.

[Greenbaum 1989], [Greenbaum, Strakoš 1992], [Paige 1976, 1980]



# A model problem

[Meurant, T. 2022]

- **Consider**  $\hat{w} = m^{-1/2}(1, \dots, 1)^T$  and  $\hat{\Lambda} = \text{diag}(\hat{\lambda}_1, \dots, \hat{\lambda}_m)$ ,

$$\hat{\lambda}_i = \hat{\lambda}_1 + \frac{i-1}{m-1}(\hat{\lambda}_m - \hat{\lambda}_1)\rho^{m-i}, \quad i = 2, \dots, m.$$

see [Strakos 1991], and “blur”  $\hat{\Lambda}$  and  $\hat{w}$ .

- **Blurred** eigenvalues and weights

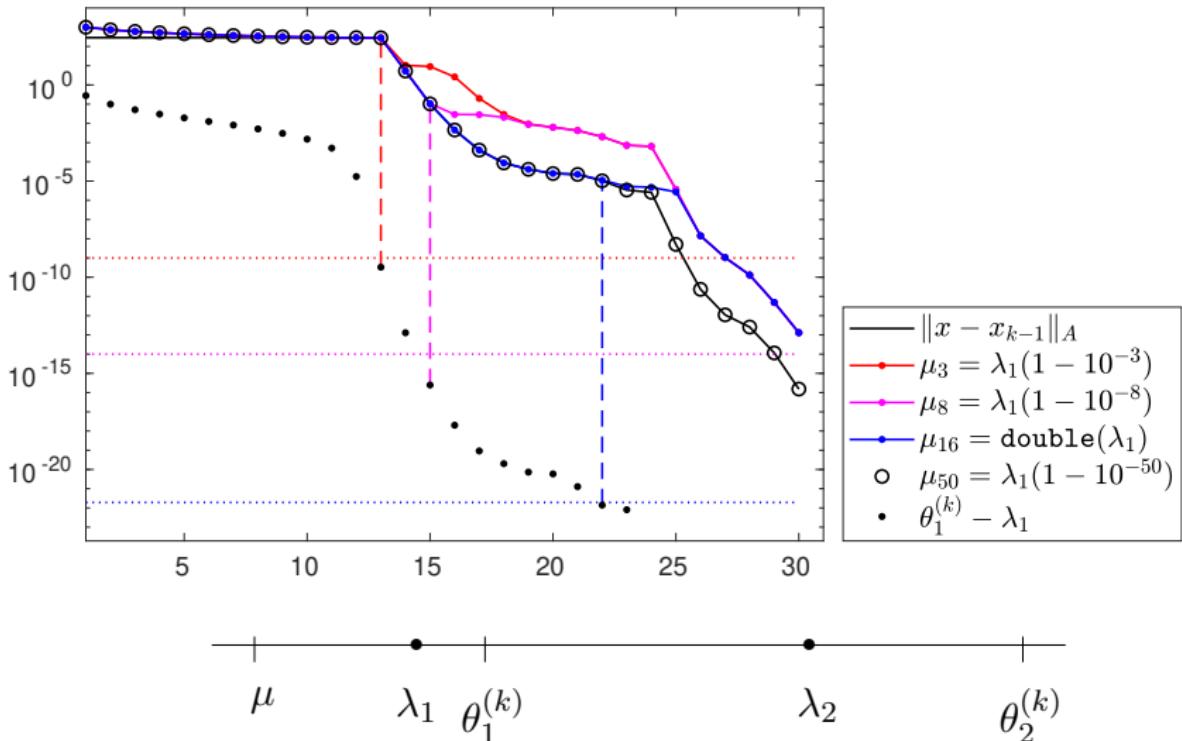
$$\lambda_j^{(i)}, \quad \omega_j^{(i)} = \frac{\hat{\omega}_i}{c_i}, \quad j = 1, \dots, c_i,$$

uniformly distributed in  $[\hat{\lambda}_i - \delta, \hat{\lambda}_i + \delta]$ ,  $c_i$  grows from 1 to  $p$ .

- $m = 12$ ,  $\hat{\lambda}_1 = 10^{-6}$ ,  $\hat{\lambda}_m = 1$ ,  $\rho = 0.8$ ,  $\delta = 10^{-10}$ ,  $p = 4$ , resulting in  $\Lambda x = w$  of **size**  $N = 30$ .

# Loss of accuracy of the Gauss-Radau upper bound

Current work [Meurant, T. 2022]



# Analysis

Based on modified tridiagonal matrices

$$T_{k+1}^{(\mu)} = \left[ \begin{array}{cc|c} \alpha_1 & \beta_1 & & \\ \beta_1 & \ddots & \ddots & \\ \ddots & \ddots & \ddots & \beta_{k-1} \\ \hline & \beta_{k-1} & \alpha_k & \beta_k \\ & \beta_k & & \alpha_{k+1}^{(\mu)} \end{array} \right]$$

that have  $\mu$  as an eigenvalue,

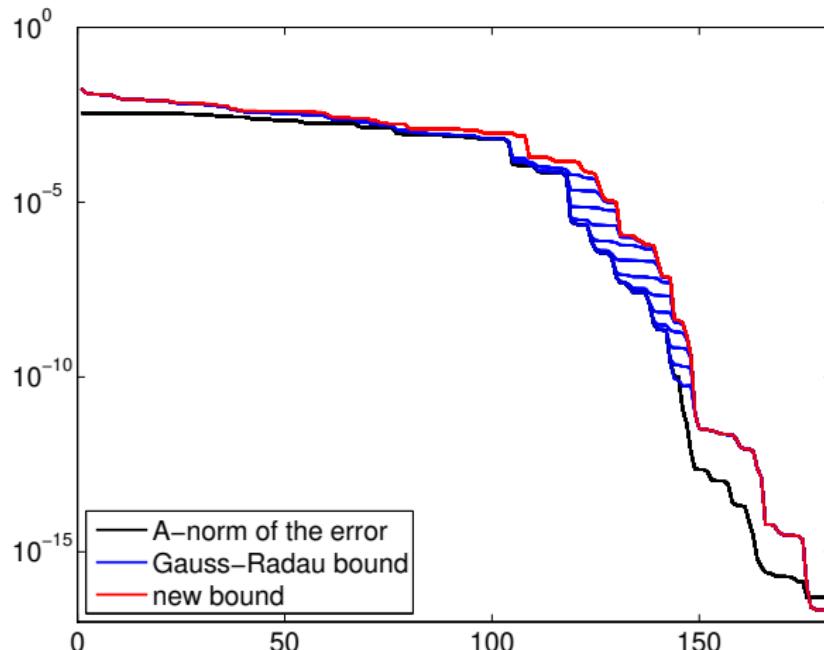
$$\alpha_{k+1}^{(\mu)} = \mu + \sum_{i=1}^k \eta_{i,k}^{(\mu)}, \quad \eta_{i,k}^{(\mu)} \equiv \frac{\left( \beta_k s_{k,i}^{(k)} \right)^2}{\theta_i^{(k)} - \mu}.$$

Analysis of behaviour of the term  $\eta_{1,k}^{(\mu)}$ , other terms are not sensitive to small modifications of  $\mu$ .

# Simple upper bound for $\mu < \lambda_{\min}$

bcsstk01,  $n = 48$ , [Meurant, T. 2019]

$$\|x - x_k\|_A^2 < \gamma_k^{(\mu)} \|r_k\|^2 < \frac{\|r_k\|^2}{\mu \|p_k\|^2} \|r_k\|^2$$



# Closeness of the Gauss-Radau upper bound and the simple upper bound

We observe that if the upper bound is delayed, then

$$\gamma_k^{(\mu)} \approx \frac{\|r_k\|^2}{\mu\|p_k\|^2}.$$

Explanation based on the formula

$$\gamma_k^{(\mu)} = \left( \frac{\mu\|p_k\|^2}{\|r_k\|^2} + \sum_{i=1}^k \left( \frac{\mu}{\theta_i^{(k)}} \right)^2 \eta_{i,k}^{(\mu)} \right)^{-1}.$$

[Meurant, T. 2022]

Improving accuracy of the upper bound

# Idea

$$\varepsilon_k = \Delta_{k:\ell-1} + \varepsilon_\ell \leq \underbrace{\Delta_{k:\ell-1} + \gamma_\ell^{(\mu)} \|r_\ell\|^2}_{\Omega_{k:\ell}^{(\mu)}}$$

Choose  $\ell$  such that

$$\frac{\Omega_{k:\ell}^{(\mu)} - \varepsilon_k}{\varepsilon_k} \leq \tau.$$

Since

$$\frac{\Omega_{k:\ell}^{(\mu)} - \varepsilon_k}{\varepsilon_k} < \frac{\Omega_{k:\ell}^{(\mu)} - \Delta_{k:\ell}}{\Delta_{k:\ell}} = \frac{\|r_\ell\|^2 (\gamma_\ell^{(\mu)} - \gamma_\ell)}{\Delta_{k:\ell}},$$

we can require

$$\frac{\|r_\ell\|^2 (\gamma_\ell^{(\mu)} - \gamma_\ell)}{\Delta_{k:\ell}} \leq \tau.$$

# Algorithm

```
1: input  $A, b, \mu, \tau$ 
2:  $r_0 = b, p_0 = r_0$ 
3:  $k = 0, \gamma_0^{(\mu)} = \frac{1}{\mu}$ 
4: for  $\ell = 0, \dots, \text{do}$ 
5:   cgiter( $\ell$ )
6:   while  $\ell \geq k$  and  $\frac{\|r_\ell\|^2 \left( \gamma_\ell^{(\mu)} - \gamma_\ell \right)}{\Delta_{k:\ell}} \leq \tau$  do
7:     accept  $\Omega_{k:\ell}^{(\mu)}$ 
8:      $k = k + 1$ 
9:   end while
10:   $\gamma_{\ell+1}^{(\mu)} = \frac{\gamma_\ell^{(\mu)} - \gamma_\ell}{\mu(\gamma_\ell^{(\mu)} - \gamma_\ell) + \delta_{\ell+1}}$ 
11: end for
```

# Conclusions

The behaviour of the Gauss-Radau upper bound

- The upper bound is **delayed** if

$$|\theta_1^{(k)} - \lambda_1| < |\mu - \lambda_1|.$$

Then, it visually coincides with the simple upper bound.

- This can happen if  $A$  has **clustered** eigenvalues (or whenever **orthogonality** is lost during computations).
- We can improve the **accuracy** of the upper bound using the information from the forthcoming CG iterations.
- The improved **Gauss lower bound** is often more **economical** than the improved Gauss-Radau upper bound.

## Related papers

G. Meurant and P. Tichý, [The behaviour of the Gauss-Radau upper bound of the error norm in CG, submitted to Numer. Algorithms.]

- G. Meurant, J. Papež, and P. Tichý, [Accurate error estimation in CG, Numer. Algorithms, 88 (2021), pp. 1337-1359.]
- G. Meurant and P. Tichý, [Approximating the extreme Ritz values and upper bounds in CG, Numer. Algorithms, 82 (2019), pp. 937-968]
- G. Meurant and P. Tichý, [On computing quadrature-based bounds for the  $A$ -norm of the error in CG, Numer. Algorithms, 62 (2013), pp. 163-191]

**Thank you for your attention!**