

Elementary Numerical Mathematics

Supporting material for

Iterative Techniques for LSE

Norms of Vectors and Matrices

Jacobi

Gauss-Seidel

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Gerhard Wellein

Department for Computer Science

HPC Services, Regionales Rechenzentrum Erlangen (RRZE)



Definition (III.1.)

A **vector norm** on \mathbb{R}^n is a function $\| \cdot \|$ from \mathbb{R}^n into \mathbb{R} with following properties:

- (1) $\|x\| \geq 0$ for all $x \in \mathbb{R}^n$
- (2) $\|x\| = 0 \Leftrightarrow x = 0$
- (3) $\|\alpha x\| = |\alpha| \|x\|$ for all $x \in \mathbb{R}^n$ and $\alpha \in \mathbb{R}$
- (4) $\|x + y\| \leq \|x\| + \|y\|$ for all $x, y \in \mathbb{R}^n$



Definition (III.2.)

The L_2 and L_∞ norm for the vector $x = (x_1, \dots, x_n)^T \in \mathbb{R}^n$ are defined by:

$$(1) \quad \|x\|_2 = \left\{ \sum_{i=1}^n x_i^2 \right\}^{1/2}$$

$$(2) \quad \|x\|_\infty = \max_{i=1, \dots, n} |x_i|$$



Definition (III.3.)

The L_2 and L_∞ distances between

$\mathbf{x} = (x_1, \dots, x_n)^T \in \mathbb{R}^n$ and $\mathbf{y} = (y_1, \dots, y_n)^T \in \mathbb{R}^n$ are defined by:

$$(1) \quad \|\mathbf{x} - \mathbf{y}\|_2 = \left\{ \sum_{i=1}^n (x_i - y_i)^2 \right\}^{1/2}$$

$$(2) \quad \|\mathbf{x} - \mathbf{y}\|_\infty = \max_{i=1, \dots, n} |x_i - y_i|$$



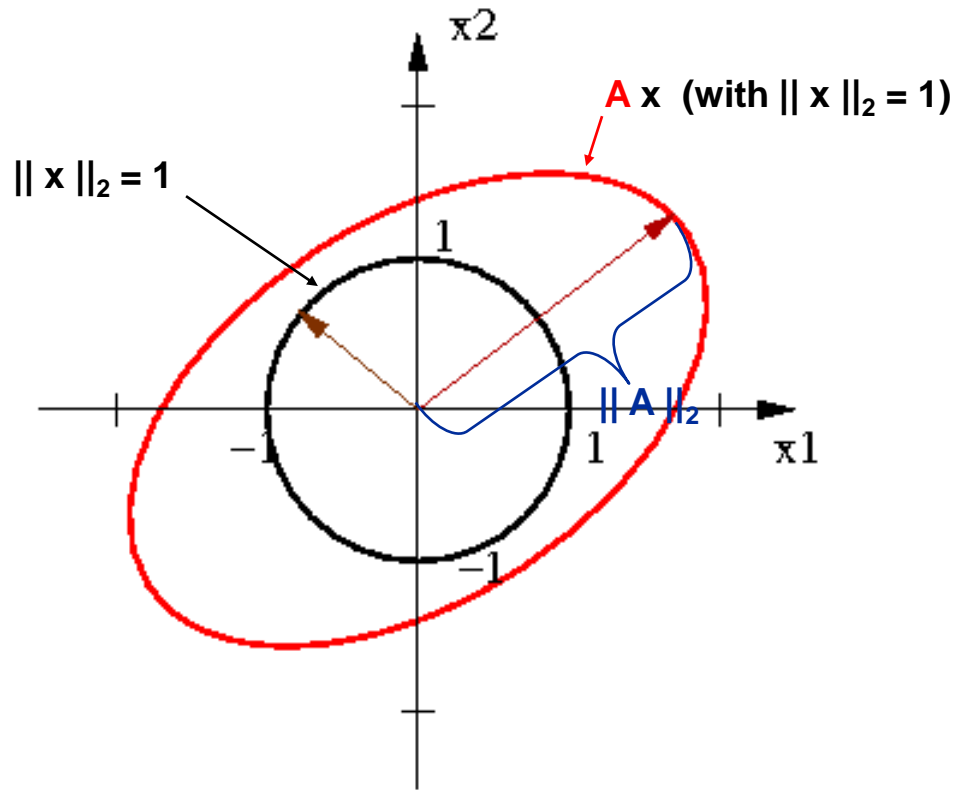
Definition (III.4.) „Cauchy-Bunyakovsky-Schwarz“

For each $x \in \mathbb{R}^n$ and $y \in \mathbb{R}^n$ the following inequality holds:

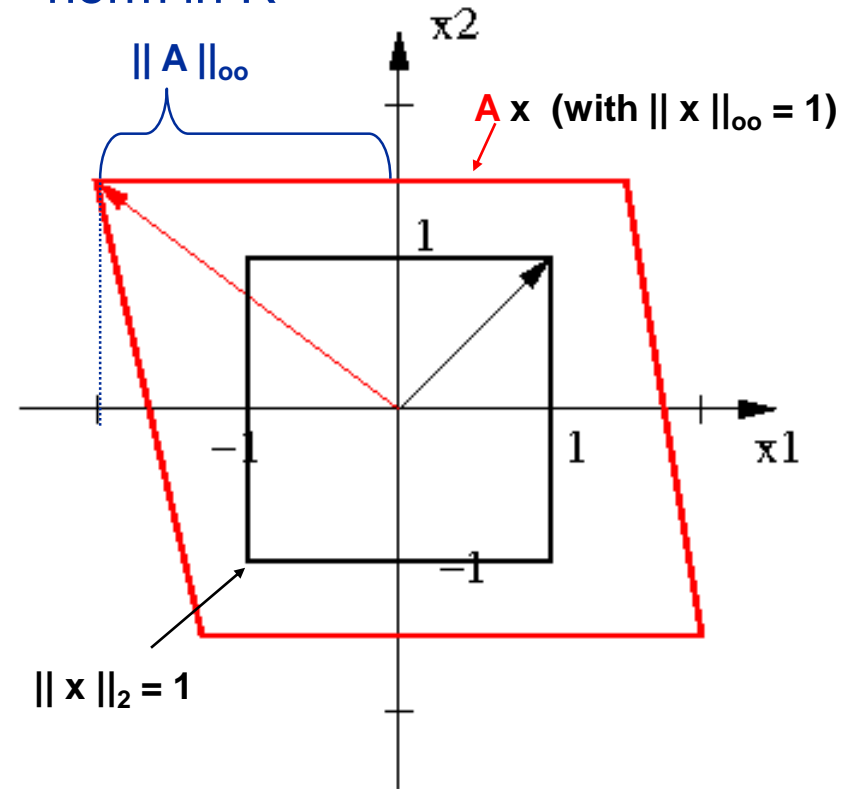
$$x^T y = \sum_{i=1}^n x_i y_i \leq \left\{ \sum_{i=1}^n x_i^2 \right\}^{\frac{1}{2}} \left\{ \sum_{i=1}^n y_i^2 \right\}^{\frac{1}{2}} = \|x\|_2 \|y\|_2$$



Matrix norm induced by L_2 vector norm in R^2



Matrix norm induced by L_{∞} vector norm in R^2



Iterative Techniques for LSE

Motivation: Jacobi method



$$E1: 10 x_1 - x_2 + 2 x_3 = 6$$

$$E2: -x_1 + 11 x_2 - x_3 + 3 x_4 = 25$$

$$E3: 2 x_1 - x_2 + 10 x_3 - x_4 = -11$$

$$E4: 3 x_2 - x_3 + 8 x_4 = 15$$

$$x_1 = (1/10) * \{ x_2 - 2 x_3 \} + (1/10) * 6$$

$$x_2 = (1/11) * \{ x_1 + x_3 - 3 x_4 \} + (1/11) * 25$$

$$x_3 = (1/10) * \{ -2 x_1 + x_2 + x_4 \} - (1/10) * 11$$

$$x_4 = (1/8) * \{ -3 x_2 + x_3 \} + (1/8) * 15$$

Define: $\mathbf{x}^{(1)} = \mathbf{T} \mathbf{x}^{(0)} + \mathbf{c}$

with: $\mathbf{c} = (6/10, 25/11, -11/10, 15/8)$

$$\mathbf{T} = \begin{pmatrix} 0 & 1/10 & -2/10 & 0 \\ 1/11 & 0 & 1/11 & -3/11 \\ -2/10 & 1/10 & 0 & 1/10 \\ 0 & -3/8 & 1/8 & 0 \end{pmatrix}$$

Iterative Techniques for LSE

Jacobi algorithm



Input: N ; $A := \text{matrix}(n, n)$; $\mathbf{x} := \text{vector}(n)$; $\mathbf{b} := \text{vector}(n)$; $\mathbf{XO} := \text{vector}(n)$

Output: $\mathbf{x} := \text{vector}(n)$; *% approximation of solution $Ax=b$*

% after k -Jacobi steps

$k=1$

while ($k \leq N$) do

for $i=1$ to n do $x_i = \frac{1}{a_{ii}} \left(- \sum_{\substack{j=1 \\ j \neq i}}^n (a_{ij} X O_j) + b_i \right)$

If ($\| \mathbf{x} - \mathbf{XO} \|$) < TOL then STOP *%(Solution reached;)*

$k = k + 1$

for $i=1$ to n do $X O_i = x_i$

$\| \cdot \|_{\infty}$ is frequently used in this context

end do

end do

end while

Iterative Techniques for LSE

Motivation: Gauss-Seidel method



$$E1: 10 x_1 - x_2 + 2 x_3 = 6$$

$$E2: -x_1 + 11 x_2 - x_3 + 3 x_4 = 25$$

$$E3: 2 x_1 - x_2 + 10 x_3 - x_4 = -11$$

$$E4: 3 x_2 - x_3 + 8 x_4 = 15$$

Use coefficients from the previous iteration

$$x_1^{(k)} = (1/10) * \{ x_2^{(k-1)} - 2 x_3^{(k-1)} + 6 \}$$

$$x_2^{(k)} = (1/11) * \{ x_1^{(k)} + x_3^{(k-1)} - 3 x_4^{(k-1)} + 25 \}$$

$$x_3^{(k)} = (1/10) * \{ -2 x_1^{(k)} + x_2^{(k)} + x_4^{(k-1)} - 11 \}$$

$$x_4^{(k)} = (1/8) * \{ -3 x_2^{(k)} + x_3^{(k)} + 15 \}$$

Use coefficients updated in the same iteration

$\{ x_1^{(k)}, x_2^{(k)}, x_3^{(k)}, x_4^{(k)} \}$ must be computed successively starting with $x_1^{(k)}$

Iterative Techniques for LSE

Gauss-Seidel algorithm



Input: N; A:= matrix(n,n); x:= vector(n); b:= vector(n); x0:= vector(n)

Output: x := vector(n); % approximation of solution Ax=b
 % after k Gauss-Seidel steps

k=1

while (k <= N) do

for i= 1 to n do

$$x_i = \frac{1}{a_{ii}} \left(- \sum_{j=1}^{i-1} a_{ij} x_j - \sum_{j=i+1}^n a_{ij} x_{0j} + b_i \right)$$

If (|| x- x0 ||) < TOL then STOP (Solution reached)

k= k + 1

for i= 1 to n do

$$x_{0i} = x_i$$

end do

end do

end while

Iterative Techniques for LSE

Jacobi – approximating the solution



	Exact solution	x_1 1,000000	x_2 2,000000	x_3 -1,000000	x_4 1,000000
<i>Initial guess</i>	0	0	0	0	0
Solution after k Jacobi steps	1	0,600000	2,272727	-1,100000	1,875000
	2	1,047273	1,715909	-0,805227	0,885227
	3	0,932636	2,053306	-1,049341	1,130881
	4	1,015199	1,953696	-0,968109	0,973843
	5	0,988991	2,011415	-1,010286	1,021351
	6	1,003199	1,992241	-0,994522	0,994434
	7	0,998128	2,002307	-1,001972	1,003594
	8	1,000625	1,998670	-0,999036	0,998888
	9	0,999674	2,000448	-1,000369	1,000619
	10	1,000119	1,999768	-0,999828	0,999786
	11	0,999942	2,000085	-1,000068	1,000109
	12	1,000022	1,999959	-0,999969	0,999960
	13	0,999990	2,000016	-1,000013	1,000019

Iterative Techniques for LSE

Gauss-Seidel – approximating the solution



	Exact solution	x_1	x_2	x_3	x_4
		1,000000	2,000000	-1,000000	1,000000
<i>Initial guess</i>		0	0	0	0
Solution after k Gauss-Seidel steps	1	0,600000	2,327273	-0,987273	0,878864
	2	1,030182	2,036938	-1,014456	0,984341
	3	1,006585	2,003555	-1,002527	0,998351
	4	1,000861	2,000298	-1,000307	0,999850
	5	1,000091	2,000021	-1,000031	0,999988
	6	1,000008	2,000001	-1,000003	0,999999
	7	1,000001	2,000000	-1,000000	1,000000
	8	1,000000	2,000000	-1,000000	1,000000



```
Input:  N; TOL,w  
        A:= matrix(n,n); x:= vector(n); b:= vector(n); x0:= vector(n)
```

```
Output: x := vector(n);           % approximation of solution Ax=b  
                                           % after k Gauss-Seidel steps
```

```
k=1
```

```
while (k <= N) do
```

```
  for i= 1 to n do  $x_i = (1 - \omega)XO_j + \frac{\omega}{a_{ii}} \left( - \sum_{j=1}^{i-1} a_{ij} x_j - \sum_{j=i+1}^n a_{ij} XO_j + b_i \right)$ 
```

```
  If ( || x- x0 || ) < TOL then STOP (Solution reached)
```

```
  k= k + 1
```

```
  for i= 1 to n do  $XO_i = x_i$ 
```

```
end while
```



Theorem III.37.

For any $x, y, z \in \mathbb{R}^n$ and $\alpha \in \mathbb{R}$ we have

1. $\langle x, y \rangle = \langle y, x \rangle$
2. $\langle \alpha x, y \rangle = \langle x, \alpha y \rangle = \alpha \langle x, y \rangle$
3. $\langle x + z, y \rangle = \langle x, y \rangle + \langle z, y \rangle$
4. $\langle x, x \rangle \geq 0$
5. $\langle x, x \rangle = 0 \iff x = 0$

Example:

$$\langle x, y \rangle := \sum_{i=1}^n x_i y_i$$



1. Choose initial approximation $x^{(0)}$
2. Compute: $r^{(0)} = b - A x^{(0)}$ and $v^{(1)} = r^{(0)}$
3. For $k = 1, 2, \dots, n$ do

a)
$$t_k = \frac{\langle r^{(k-1)}, r^{(k-1)} \rangle}{\langle v^{(k)}, A v^{(k)} \rangle}$$

b)
$$x^{(k)} = x^{(k-1)} + t_k v^{(k)}$$

! Calc. new approximation

c)
$$r^{(k)} = r^{(k-1)} - t_k A v^{(k)}$$

! Calc. current residual

d)
$$s_k = \frac{\langle r^{(k)}, r^{(k)} \rangle}{\langle r^{(k-1)}, r^{(k-1)} \rangle}$$

e)
$$v^{(k+1)} = r^{(k)} + s_k v^{(k)}$$

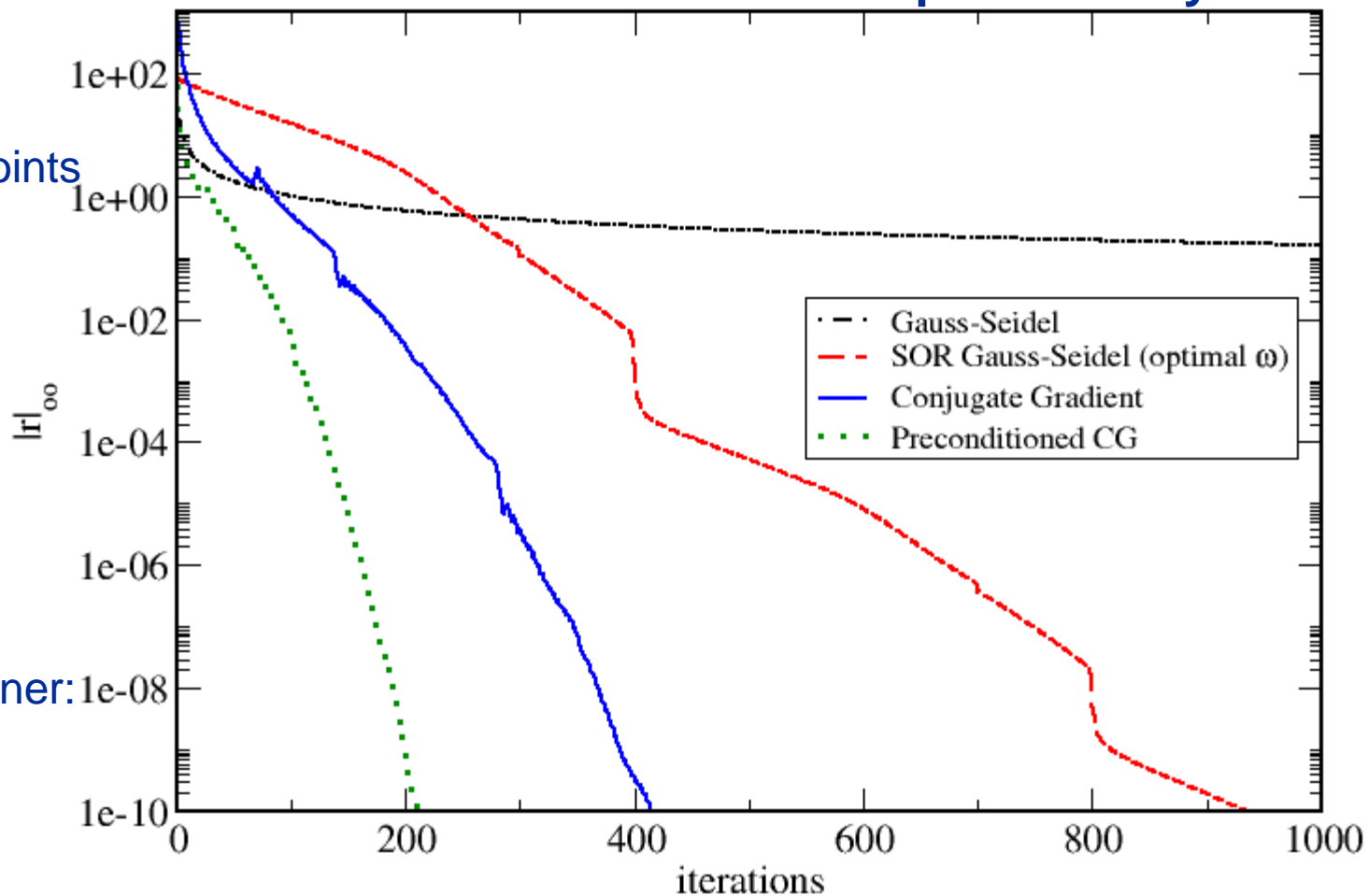
! Calc. Next search direction

f) Check for convergence of $x^{(k)}$



- **Solve 2D Poisson equation ($\Delta u(x,y)=f(x,y)$) using finite difference method. Coefficient matrix has 5 nonzero elements per row only!**

- 200*100 grid points
→ $n=20.000$



- CG Preconditioner: Symmetr. GS

- Compute times ~ few seconds; Gaussian elimination: hours (estimate)



If the spectral radius $\rho(T)$ satisfies $\rho(T) < 1$,

then $(I - T)^{-1}$ exists and

$$(I - T)^{-1} = I + T + T^2 + \dots = \sum_{j=0}^{\infty} T^j$$



- **On travel Nov 16th – 22nd → No lectures in that week!**
- **Lecture Nov. 18th → Nov 13th H4 18:00 (s.t.)**
- **Lecture Nov. 21st → Nov 26th H4 18:00 (s.t.)**