

Scientific Computing

Lecture 6

Delft University of Technology

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Today

- Remaining Chapter 4:
 - Partial Pivoting
 - Bandwidth
 - Cholesky
- Chapter 5:
 - Basic Iterative Methods + convergence
 - Richardson Method

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THE HURICANE IS MOVING TOWARD THE NORTHEAST OR 050 DEGREES AT 21 KNOTS. SINCE THE HURRICANE IS ALREADY EMBEDDED WITHIN THE MID-LATITUDE WESTERLIES....IT SHOULD CONTINUE ON THIS GENERAL TRACK WITH AN INCREASE IN FORWARD SPEED FOR THE NEXT FEW DAYS.

NO 96-HOUR POINT IS BEING GIVEN BECAUSE FORECAST POINTS IN THE EASTERN HEMISPHERE BREAK A LOT OF SOFTWARE.

FORECAST POSITIONS AND MAX WINDS

Recap

LU Decomposition for General Matrices

$A \in \mathbb{R}^{n \times n}$ A is non singular

$$A^{(k-1)} = M_{k-1} M_{k-2} \cdot M_1 A = \begin{pmatrix} A_{11}^{(k-1)} & A_{12}^{(k-1)} \\ 0 & A_{22}^{(k-1)} \end{pmatrix}$$

M_k is Gauss transformation: $M_k = I - a^{(k)} e_k^T$

$$e_k = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} \quad \text{and} \quad a^{(k)} = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ a_{k+1,k}^{(k-1)} / a_{k,k}^{(k-1)} \\ \vdots \\ a_{n,k}^{(k-1)} / a_{k,k}^{(k-1)} \end{pmatrix}$$

$a^{(k)}$ is the Gauss vector and we must have pivot $a_{k,k}^{(k-1)} \neq 0$.

$$M_k = I - a^{(k)} e_k^T$$

Recap

LU Decomposition for General Matrices

Gaussian elimination gives:

$$Au = f$$

$$M_1Au = M_1f$$

$$\vdots$$

$$M_{n-1} \dots M_1Au = M_{n-1} \dots M_1f$$

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$$M_{n-1} \dots M_1 Au = M_{n-1} \dots M_1 f$$

$$M_{n-1} \dots M_1 Au = Uu = M_{n-1} \dots M_1 f \Rightarrow \underbrace{M_1^{-1} M_2^{-1} \dots M_{n-1}^{-1}}_{\text{is this } L?} Uu = f$$

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LU Decomposition for General Matrices

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$$M_{n-1} \dots M_1 Au = Uu = M_{n-1} \dots M_1 f \Rightarrow \underbrace{M_1^{-1} M_2^{-1} \dots M_{n-1}^{-1}}_{\text{is this } L?} Uu = f$$

Yes! See Lemma 4.4.2, which states

$$\begin{aligned} (M_{n-1} \dots M_1)^{-1} &= M_1^{-1} \dots M_{n-1}^{-1} \\ &= \prod_{k=1}^{n-1} (I + \alpha^{(k)} \mathbf{e}_k^T) \\ &= I + \sum_{k=1}^{n-1} \alpha^{(k)} \mathbf{e}_k^T \end{aligned}$$

Example: LU w/o Pivoting (from *Numerical Linear Algebra*)

$$A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

Step $k = 1$: select pivot $a_{11} = 2$ and construct $a^{(1)}$

$$a^{(1)} = \begin{pmatrix} 0 \\ \frac{a_{12}}{a_{11}} \\ \frac{a_{21}}{a_{11}} \\ \frac{a_{13}}{a_{11}} \\ \frac{a_{11}}{a_{11}} \\ \frac{a_{14}}{a_{11}} \\ \frac{a_{11}}{a_{11}} \end{pmatrix} = \begin{pmatrix} 0 \\ 2 \\ 4 \\ 4 \\ 3 \end{pmatrix}, \quad M_1 = I - a^{(1)}e_1^T = \begin{bmatrix} 1 & & & \\ -2 & 1 & & \\ -4 & & 1 & \\ -3 & & & 1 \end{bmatrix}$$

$$M_1 A = \begin{bmatrix} 1 & & & \\ -2 & 1 & & \\ -4 & & 1 & \\ -3 & & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 1 & 1 & 1 & 1 \\ 3 & 5 & 5 & 5 \\ 4 & 6 & 8 & 8 \end{bmatrix}$$

Example: LU w/o Pivoting

$$M_1 A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & 3 & 5 & 5 \\ & 4 & 6 & 8 \end{bmatrix}$$

Step $k = 2$: select new pivot $a_{22} = 1$ and construct $a^{(2)}$

$$a^{(2)} = \begin{pmatrix} 0 \\ 0 \\ \frac{a_{32}}{a_{22}} \\ \frac{a_{42}}{a_{22}} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 3 \\ 4 \end{pmatrix}, \quad M_2 = I - a^{(2)} e_2^T = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & -3 & 1 & \\ & -4 & & 1 \end{bmatrix}$$

$$M_2(M_1 A) = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & -3 & 1 & \\ & -4 & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & 3 & 5 & 5 \\ & 4 & 6 & 8 \end{bmatrix} = \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & 2 & 4 \end{bmatrix}$$

Example: LU w/o Pivoting

$$M_2(M_1A) = \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & & 2 & 4 \end{bmatrix}$$

Step $k = 3$: select new pivot $a_{33} = 2$ and construct $a^{(3)}$

$$a^{(3)} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \frac{a_{43}}{a_{33}} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix}, \quad M_3 = I - a^{(3)}e_3^\top = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & -1 & 1 \end{bmatrix}$$

$$M_3(M_2M_1A) = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & & 2 & 4 \end{bmatrix} =$$

$$\begin{bmatrix} 2 & 1 & 1 & 0 \\ & 1 & 1 & 1 \\ & & 2 & 2 \\ & & & 2 \end{bmatrix} = U$$

Recap

Example: LU Decomposition for General Matrices (from video lecture)

We start with A , $k = 2$, so $k = 1$ step done. Red block is $A_{2,2}^{(1)}$.

$$A^{(k-1)} = A^{(1)} = \begin{pmatrix} 3 & 2 & 0 \\ 0 & 2 & -1 \\ 0 & -1 & 2 \end{pmatrix}$$

Pivot is $a_{2,2}^{(1)} = 2$, so Gauss vector becomes $a^{(2)} = \begin{pmatrix} 0 \\ 0 \\ -\frac{1}{2} \end{pmatrix}$

$$M_2 = I - a^{(2)} e_1^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & \frac{1}{2} & 1 \end{pmatrix}$$

Recap

Example: LU Decomposition for General Matrices

$$M_2 = I - a^{(2)}e_1^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & \frac{1}{2} & 1 \end{pmatrix}$$

This gives:

$$M_2 A^{(1)} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & \frac{1}{2} & 1 \end{pmatrix} \begin{pmatrix} 3 & 2 & 0 \\ 0 & 2 & -1 \\ 0 & -1 & 2 \end{pmatrix} = \begin{pmatrix} 3 & 2 & 0 \\ 0 & 2 & -1 \\ 0 & 0 & \frac{3}{2} \end{pmatrix} = U$$

Recap

§ 4.5.2: Rounding Errors

$$\hat{L}, \hat{U} \Rightarrow \hat{L} \cdot \hat{U} = A + E, \|E\| \leq n\mu \|\hat{L}\| \|\hat{U}\|$$

Recap

§ 4.5.2: Rounding Errors

$$\hat{L}, \hat{U} \Rightarrow \hat{L} \cdot \hat{U} = A + E, \|E\| \leq n\mu \|\hat{L}\| \|\hat{U}\|$$

Example

Suppose we want to solve $Au = f$ on a computer which rounds to 3 digits:

$$\begin{pmatrix} 0.001 & 1 \\ 1 & -4 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 1 \\ -3 \end{pmatrix}$$

Recap

§ 4.5.2: Rounding Errors

$$\hat{L}, \hat{U} \Rightarrow \hat{L} \cdot \hat{U} = A + E, \|E\| \leq n\mu \|\hat{L}\| \|\hat{U}\|$$

Example

Suppose we want to solve $Au = f$ on a computer which rounds to 3 digits:

$$\begin{pmatrix} 0.001 & 1 \\ 1 & -4 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 1 \\ -3 \end{pmatrix}$$

Well conditioned problem but due to your LU decomposition becomes unstable.

Pivoting: Introduction

Suppose we want to solve $Au = f$ on a computer which rounds to 3 digits:

$$\begin{pmatrix} 0.001 & 1 \\ 1 & -4 \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \end{pmatrix} = \begin{pmatrix} 1 \\ -3 \end{pmatrix}$$

$$u = \begin{pmatrix} 1 \\ 1 \end{pmatrix} \Rightarrow \hat{u} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} = u + \Delta u$$

§ 4.6: Pivoting

Gaussian Elimination with Partial Pivoting

$$A = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ a_{21} & & \\ \vdots & & \\ a_{n1} & \cdots & a_{nn} \end{pmatrix}$$

Determine \hat{j} such that $|a_{\hat{j},1}| \geq |a_{j,1}|$ for $1 \leq j \leq n$.

Next, change row 1 with row \hat{j} :

$$\begin{pmatrix} a_{\hat{j}1} & \cdots & \cdots & a_{\hat{j}n} \\ a_{21} & \cdots & \cdots & a_{2n} \\ a_{11} & \cdots & \cdots & a_{1n} \\ \vdots & & & \vdots \\ a_{n1} & & & a_{nn} \end{pmatrix}$$

§ 4.6: Pivoting

Gaussian Elimination with Partial Pivoting

Linear system to solve becomes:

$$\begin{pmatrix} a_{j1} & \cdots & \cdots & a_{jn} \\ a_{21} & \cdots & \cdots & a_{2n} \\ a_{11} & \cdots & \cdots & a_{1n} \\ \vdots & & & \vdots \\ a_{n1} & & & a_{nn} \end{pmatrix} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \\ \vdots \\ u_n \end{pmatrix} = \begin{pmatrix} f_j \\ f_2 \\ f_1 \\ \vdots \\ f_n \end{pmatrix}$$

Note: new multiplier becomes $\frac{|a_{i,1}|}{|a_{j,1}|}$ for $1 \leq i \leq n$.

We can show at the end that $PA = LU$, which leads to solving:

$$Au = f \quad \text{solve } y \text{ from } Ly = Pf$$

$$PAu = Pf \quad \text{solve } u \text{ from } Uu = y$$

$$L \underbrace{Uu}_y = Pf$$

Example: LU with Pivoting

$$A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

Step $k = 1$: we start with selecting the largest value in the first column (in original matrix A , element $a_{13} = 8$).

That will be our pivot so we switch rows 1 and 3 by multiplying with the permutation matrix P_1 :

$$\begin{bmatrix} & & 1 & \\ & 1 & & \\ 1 & & & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ 4 & 3 & 3 & 1 \\ 8 & 7 & 9 & 5 \\ 6 & 7 & 9 & 8 \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

P_1

Example: LU with Pivoting

$$P_1 A = \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix}$$

Step $k = 1$: select pivot new $a_{11} = 8$ and construct $a^{(1)}$

$$a^{(1)} = \begin{pmatrix} 0 \\ \frac{a_{12}}{a_{11}} \\ \frac{a_{21}}{a_{11}} \\ \frac{a_{31}}{a_{11}} \\ \frac{a_{41}}{a_{11}} \end{pmatrix} = \begin{pmatrix} 0 \\ \frac{1}{2} \\ \frac{1}{4} \\ \frac{3}{4} \end{pmatrix}, \quad M_1 = I - a^{(1)} e_1^T = \begin{bmatrix} 1 & & & \\ -\frac{1}{2} & 1 & & \\ -\frac{1}{4} & & 1 & \\ -\frac{3}{4} & & & 1 \end{bmatrix}$$

$$M_1(P_1 A) = \begin{bmatrix} 1 & & & \\ -\frac{1}{2} & 1 & & \\ -\frac{1}{4} & & 1 & \\ -\frac{3}{4} & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ 4 & 3 & 3 & 1 \\ 2 & 1 & 1 & 0 \\ 6 & 7 & 9 & 8 \end{bmatrix} =$$

$$\begin{bmatrix} 8 & 7 & 9 & 5 \\ -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} & -\frac{3}{2} \\ -\frac{1}{4} & -\frac{5}{4} & -\frac{5}{4} & -\frac{5}{4} \\ -\frac{3}{4} & -\frac{9}{4} & -\frac{17}{4} & -\frac{17}{4} \end{bmatrix}$$

Example: LU with Pivoting

$$P_2(M_1 P_1 A) = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix}$$

Step $k = 2$: select pivot new $a_{22} = \frac{7}{4}$ and construct $a^{(2)}$

$$a^{(2)} = \begin{pmatrix} 0 \\ 0 \\ \frac{a_{32}}{a_{22}} \\ \frac{a_{42}}{a_{22}} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ -\frac{3}{7} \\ -\frac{7}{7} \end{pmatrix}, \quad M_2 = I - a^{(2)} e_2^T = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & \frac{3}{7} & 1 & \\ & \frac{7}{7} & & 1 \end{bmatrix}$$

$$M_2(P_2 M_1 P_1 A) = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & \frac{3}{7} & 1 & \\ & \frac{7}{7} & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{3}{4} & -\frac{5}{4} & -\frac{5}{4} \\ & -\frac{1}{2} & -\frac{3}{2} & -\frac{3}{2} \end{bmatrix} =$$

$$\begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & -\frac{2}{7} & -\frac{4}{7} & -\frac{6}{7} \\ & -\frac{6}{7} & -\frac{2}{7} & -\frac{2}{7} \end{bmatrix}$$

Example: LU with Pivoting

$$M_2(P_2M_1P_1A) = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{2}{7} & \frac{4}{7} \\ & & -\frac{6}{7} & -\frac{2}{7} \end{bmatrix}$$

Step $k = 3$: interchange rows 3 and 4, by multiplying by P_3 as largest value in magnitude is $a_{34} = -\frac{6}{7}$:

$$\begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{2}{7} & \frac{4}{7} \\ & & -\frac{6}{7} & -\frac{2}{7} \end{bmatrix} = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & \frac{7}{4} & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & -\frac{2}{7} & \frac{4}{7} \end{bmatrix}$$

P_3

Example: LU with Pivoting

$$P_3 M_2 P_2 M_1 P_1 A = \begin{bmatrix} 8 & 7 & 9 & 5 \\ & 7 & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & -\frac{2}{7} & \frac{4}{7} \end{bmatrix}$$

Step $k = 3$: select new pivot $a_{33} = -\frac{6}{7}$ and construct $a^{(3)}$

$$a^{(2)} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \frac{a_{43}}{a_{33}} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ \frac{1}{3} \end{pmatrix}, \quad M_3 = I - a^{(3)} e_3^T = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & -\frac{1}{3} & 1 \end{bmatrix}$$

$$M_3(P_3 M_2 P_2 M_1 P_1 A) = \begin{bmatrix} 1 & & & \\ & 1 & & \\ & & 1 & \\ & & -\frac{1}{3} & 1 \end{bmatrix} \begin{bmatrix} 8 & 7 & 9 & 5 \\ & 7 & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & -\frac{2}{7} & \frac{4}{7} \end{bmatrix} =$$

$$\begin{bmatrix} 8 & 7 & 9 & 5 \\ & 7 & \frac{9}{4} & \frac{17}{4} \\ & & -\frac{6}{7} & -\frac{2}{7} \\ & & & \frac{2}{3} \end{bmatrix} = U$$

§ 4.8: Diagonal Dominance and Pivoting

Is pivoting necessary?

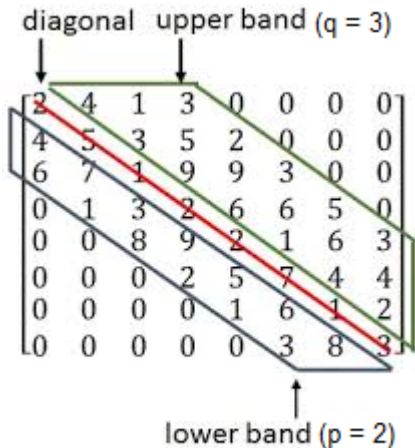
If A is strictly diagonal dominant $|a_{ii}| > \sum_{j=1, j \neq i}^n |a_{j,i}|$, then partial pivoting is not needed.

§ 4.9: Cholesky Decomposition

If A SPD, then $A = CC^T$:

- Symmetric: $A = A^T$
- Positive Definite: $x^T Ax > 0 \quad x \in \mathbb{R}^1, x \neq 0$

§ 4.10: Bandwidth



Pivoting destroys the 'nice' bandwidth

Industrial application: unstructured grids, band structure also gone (§ 4.11).

§ 5.2: Basic Iterative Methods (Introduction)

We assume that A is an invertible matrix so A^{-1} exists.
Iterative methods seek solutions to:

$$Au = f$$

by forming the sequences of iterates by

$$\{u^k\}_{k \geq 0} \text{ where } u^k \rightarrow u \text{ for } k \rightarrow \infty$$

where u_0 is the initial guess.

§ 5.3: Basic Iterative Methods

Notation: u^k is the k-th iteration. We define:

- Error: $e^k = u - u^k$
- Residual: $r^k = f - Au^k$
- Residual equation: $Ae^k = Au - Au^k = f - Au^k = r^k$

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We start with a **splitting method**: $A = M - N$.

$$Au = f \Rightarrow (M - N)u = f = Mu = f + Nu$$

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$$Mu^{k+1} = f + Nu^k$$

$$u^{k+1} = M^{-1}f + M^{-1}Nu^k$$

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$$Au = f \Rightarrow (M - N)u = f = Mu = f + Nu$$

$$Mu^{k+1} = f + Nu^k$$

$$u^{k+1} = M^{-1}f + M^{-1}Nu^k$$

Substituting $n = M - A$ we get:

$$\begin{aligned}u^{k+1} &= M^{-1}f + M^{-1}(M - A)u^k \\ &= u^k + M^{-1}(f - Au^k) \\ &= u^k + M^{-1}(r^k)\end{aligned}$$

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- Error: $e^k = u - u^k$
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Substituting $n = M - A$ we get:

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Note: M^{-1} only for analysis!

§ 5.4.1: Convergence

Does this iterative scheme converge? I.e. $u^{k+1} \rightarrow u$ as k goes to infinity?

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Does this iterative scheme converge? I.e. $u^{k+1} \rightarrow u$ as k goes to infinity?

$$u^{k+1} = u^k + M^{-1} (f - Au^k)$$

$$u - u^{k+1} = u - u^k - M^{-1} (Au - Au^k)$$

$$e^{k+1} = e^k - M^{-1} A e^k \Rightarrow e^{k+1} = (I - M^{-1} A) e^k$$

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Does this iterative scheme converge? I.e. $u^{k+1} \rightarrow u$ as k goes to infinity?

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$$u - u^{k+1} = u - u^k - M^{-1} (Au - Au^k)$$

$$e^{k+1} = e^k - M^{-1} A e^k \Rightarrow e^{k+1} = (I - M^{-1} A) e^k$$

We define the **iteration matrix** $B = I - M^{-1} A$ such that:

$$e^{k+1} = B^k e^0$$

§ 5.4.1: Convergence

Does this iterative scheme converge? I.e. $u^{k+1} \rightarrow u$ as k goes to infinity?

$$u^{k+1} = u^k + M^{-1} (f - Au^k)$$

$$u - u^{k+1} = u - u^k - M^{-1} (Au - Au^k)$$

$$e^{k+1} = e^k - M^{-1} A e^k \Rightarrow e^{k+1} = (I - M^{-1} A) e^k$$

We define the **iteration matrix** $B = I - M^{-1}A$ such that:

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For convergence: $\lim_{k \rightarrow \infty} \|B^k\|_2 = 0$, which using theorem 2.7.2 is equivalent to:

$$\lim_{k \rightarrow \infty} \|B^k\|_2 = 0 \Leftrightarrow \rho(B) < 1$$

§ 5.3.2: Richardson Method

One simple method is the **Richardson Method**: $M = I, N = I - A$, then $u^{k+1} = f + (I - A)u^k$

Hence, iteration matrix $B = I - M^{-1}A = I - A$.

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Hence, iteration matrix $B = I - M^{-1}A = I - A$.

$$u^0 = 0$$

$$u^1 = f$$

$$u^2 = f + (I - A)f$$

$$u^3 = f + (I - A)f + (I - A)^2f$$

\vdots

$$u^{k+1} = \sum_{i=0}^k (I - A)^i f$$

Does this iterative scheme converge? I.e. k to infinity and $u^{k+1} \rightarrow u$, implies the sum $\sum_{i=0}^k (I - A)^i$ should go to A^{-1} .