MA5233 Computational Mathematics

Lecture 3: LU Factorisation

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LU Factorisation

LU factorisation
Algorithm of choice for solving dense linear systems.

Outline

- LU factorisation: why and how.
- `lu()` in Julia.
- Computational complexity of LU factorisation.
- Conditioning and stability of LU factorisation.

LU Factorisation

Linear system of equations
Given $A \in \mathbb{K}^{n \times n}$ and $b \in \mathbb{K}^n$, find $x \in \mathbb{K}^n$ such that $Ax = b$.

Observation
Problem is easy if A is triangular, i.e.

$$A(i,j) = 0 \qquad \text{for} \qquad \begin{cases} i > j & \text{(upper triangular)}, \\ i < j & \text{(lower triangular)}. \end{cases}$$

LU Factorisation

Example

$$\begin{pmatrix} 4 & 1 & -2 \\ 0 & 2 & -1 \\ 0 & 0 & 3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3 \\ 4 \\ 6 \end{pmatrix}$$

$$\begin{aligned}
&\text{Third eq.:} & 3x_3 = 6 &\implies x_3 = 2 \\
&\text{Second eq.:} & 2x_2 - x_3 = 4 &\implies x_2 = 3 \\
&\text{First eq.:} & 4x_1 + x_2 - 2x_3 = 3 &\implies x_1 = 1
\end{aligned}$$

LU Factorisation

LU factorisation theorem
For every invertible matrix $A \in \mathbb{K}^{n \times n}$, there exist

- a permutation matrix $P \in \mathbb{K}^{n \times n}$,
- a lower-triangular matrix $L \in \mathbb{K}^{n \times n}$ with unit diagonal, and
- an upper-triangular matrix $U \in \mathbb{K}^{n \times n}$

such that $PA = LU$.

L, U are unique for fixed P.

Why is there a P in this theorem?

Because of cases like this one:

$$\begin{pmatrix} 0 & 1 \\ 1 & 1 \end{pmatrix}$$

LU Factorisation

Example

$$\begin{pmatrix} 4 & 1 & -2 \\ -8 & 0 & 3 \\ 12 & 7 & -5 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 3 & 0 & 1 \end{pmatrix} \begin{pmatrix} 4 & 1 & -2 \\ 0 & 2 & -1 \\ 0 & 4 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix} \begin{pmatrix} 4 & 1 & -2 \\ 0 & 2 & -1 \\ 0 & 0 & 3 \end{pmatrix}$$

LU Factorisation

Column/partial pivoting

Use row with largest entry in first column to eliminate the other rows.

Example

$$\begin{pmatrix} 4 & 1 & -2 \\ -8 & 0 & 3 \\ \color{red}{12} & 7 & -5 \end{pmatrix} \to \begin{pmatrix} \color{red}{12} & 7 & -5 \\ -8 & 0 & 3 \\ 4 & 1 & -2 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ -\frac{2}{3} & 1 & 0 \\ \frac{1}{3} & 0 & 1 \end{pmatrix} \begin{pmatrix} \color{red}{12} & 7 & -5 \\ 0 & x & x \\ 0 & x & x \end{pmatrix}$$

LU Factorisation

Complete pivoting

Use largest overall entry (i, j) to eliminate the other entries in column j.

Example

$$\begin{pmatrix} 1 & 4 & -2 \\ 0 & -8 & 3 \\ 7 & \color{red}{12} & -5 \end{pmatrix} \to \begin{pmatrix} \color{red}{12} & 7 & -5 \\ -8 & 0 & 3 \\ 4 & 1 & -2 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ -\frac{2}{3} & 1 & 0 \\ \frac{1}{3} & 0 & 1 \end{pmatrix} \begin{pmatrix} \color{red}{12} & 7 & -5 \\ 0 & x & x \\ 0 & x & x \end{pmatrix}$$

LU Factorisation

Permutations
Bijective map $\pi : \{1, \dots, n\} \to \{1, \dots, n\}$.

Example

$$\pi(1) = 2, \qquad \pi(2) = 4, \qquad \pi(3) = 3, \qquad \pi(4) = 1$$

Representations

$$P = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \qquad \text{or} \qquad p = \begin{pmatrix} 2 & 4 & 3 & 1 \end{pmatrix}^T$$

LU Factorisation

Remarks on permutations

- PA permutes rows of A.
- Matrix representation is convenient, but very inefficient.
  Never use permutation matrices in your code!
- Applying permutation `p` in Julia:
  ```
  data = ["a","b","c","d"]
  p = [2,4,3,1]
  data[p] -> ["b","d","c","a"]
  ```
- Permutations are by definition bidirectional.
  Be careful which direction you represent in your code!

LU Factorisation

Solving $Ax = b$ via LU factorisation

- Compute LU factorisation $PA = LU$.
- Permute the RHS: $\hat{b} = P^{-1}b$.
- Solve $y = L^{-1}\hat{b}$.
- Solve $x = U^{-1}y$.

LU Factorisation

Solving $Ax = b$ in Julia

`F = lu(A, pivot=Val(true))` computes "factorisation object".

- Access factors through `F.L`, `F.U`, `F.p` (vector) and `F.P` (matrix).
- `F.L * F.U == A[F.p, :]  == F.P * A`.
- `x = F\b` computes solution.
- `A\b` solves `A*x = b` directly.

LU Factorisation

Measuring the "required effort" of an algorithm

Some ideas:

- Count the number of `+,-,x,/,sqrt`.

  Very tedious, and makes it hard to compare algorithms.

- Measure its runtime.

  Too dependent on input, hardware, etc.

Most common measure: big-$\mathcal{O}$ estimate.

Examples

- Evaluating $x^T y := \sum_{k=1}^{n} x_i y_i$ takes
  - n multiplications, and
  - $n-1$ additions.

  Computing inner products takes $\mathcal{O}(n)$ floating-point operations.

- Ax for $A \in \mathbb{K}^{n\times n}$ can be computed as n inner products.
  Evaluating matvec takes $\mathcal{O}(n^2)$ floating-point operations.

LU Factorisation

Why is big-$\mathcal{O}$ notation useful?
It tells us the functional dependency of runtime on problem size.

$$\begin{aligned} \mathcal{O}(n) \text{ algorithm} &\quad\Longrightarrow\quad \text{Changing } n \to 2n \text{ multiplies runtime by 2.} \\ \mathcal{O}(n^2) \text{ algorithm} &\quad\Longrightarrow\quad \text{Changing } n \to 2n \text{ multiplies runtime by 4.} \\ &\quad\ldots \end{aligned}$$

Computational cost of LU factorisation

- Factorisation: $\mathcal{O}(n^3)$.
- Triangular solves: $\mathcal{O}(n^2)$.

Hence, reuse factorisation if possible.

LU Factorisation

Conditioning of linear systems

Assume

- $Ax = b$, and
- $(A + \Delta A)(x + \Delta x) = b + \Delta b$ with $\|\Delta A\| < \|A^{-1}\|^{-1}$.

Then,

$$\begin{aligned}\frac{\|\Delta x\|}{\|x\|} &\leq \frac{\kappa(A)}{1 - \kappa(A)\frac{\|\Delta A\|}{\|A\|}} \left(\frac{\|\Delta A\|}{\|A\|} + \frac{\|\Delta b\|}{\|b\|}\right) \\ &\approx \kappa(A) \left(\frac{\|\Delta A\|}{\|A\|} + \frac{\|\Delta b\|}{\|b\|}\right) + \mathcal{O}\big(\kappa(A)^2\big).\end{aligned}$$

Statement to remember

Relative error in x is at most $\kappa(A)$ times relative errors in A and b up to higher-order terms.

LU Factorisation

Stability of solving linear systems via LU factorisation
Numerical solution $\tilde{x}$ to $Ax = b$ computed via LU factorisation satisfies

$$(A + \Delta A)\,\tilde{x} = b \qquad \text{where} \qquad \frac{\|\Delta A\|}{\|L\|\,\|U\|} \approx \mathcal{O}\big(\varepsilon_{\mathsf{mach}}\big).$$

Combined with condition number for linear systems, this yields

$$\frac{\|\tilde{x} - x\|}{\|x\|} \approx \kappa(A)\,\frac{\|L\|\,\|U\|}{\|A\|}\,\mathcal{O}\big(\varepsilon_{\mathsf{mach}}\big).$$

Error is small if

- $\kappa(A)$ is not too large (problem is well-conditioned), and
- $\frac{\|L\|\,\|U\|}{\|A\|}$ is not too large (LU factorisation is stable).

LU Factorisation

Impact of pivoting

No pivoting: $\|L\|, \|U\| = \infty$ is possible.

- We will see special matrices which do not require pivoting.
- Do not use this algorithm unless you know what you are doing.

Partial pivoting: $\|L\|, \|U\| \leq 2^{n-1}$ is a sharp upper bound.

- However, exponential growth of $\|L\|, \|U\|$ has never been observed in practice.
- Famous quote: "Anyone that unlucky has already been run over by a bus."
- This is the recommended algorithm in most applications.

Complete pivoting: probably $\|L\|, \|U\| = \mathcal{O}(n)$.

- No one uses this algorithm.

LU Factorisation

References and further reading

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