

- ▶ **Gaussian elimination**
- ▶ **Matrix multiplication**
- ▶ **LU factorization**
- ▶ **Inverse matrix and Gauss-Jordan elimination**
- ▶ **Ill-conditioned matrices and round-off errors**

SUGGESTED READING:

G. Strang, Linear algebra and its applications, Chapter 1

Gaussian Elimination, Geometry, Mechanics & Cost

- ▶ Reduce $N \times N$ system by repeatedly subtracting multiples of one equation from another equation
- ▶ Geometry: intersection of n subspaces (“hyper-planes”), singular cases
- ▶ Mechanics: forward elimination (clear out columns below pivots, $A \rightarrow U$) and back-substitution. Stop and think when zero pivot encountered (Row exchange? Look *below the zero pivot* for non-zero entry. Failure signifies singularity, i.e. no solution or ∞ of solutions.)
- ▶ Cost: FE $\approx \frac{1}{3}n^3$, BS $\approx \frac{1}{2}n^2$.

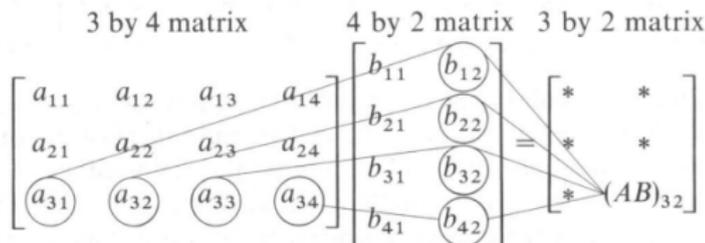
EXAMPLE

$[A|b] \rightarrow [U|b]$:

$$\begin{bmatrix} 2 & 1 & 1 & 5 \\ 4 & -6 & 0 & -2 \\ -2 & 7 & 2 & 9 \end{bmatrix} \rightarrow \begin{bmatrix} 2 & 1 & 1 & 5 \\ 0 & -8 & -2 & -12 \\ 0 & 8 & 3 & 14 \end{bmatrix} \rightarrow \begin{bmatrix} \boxed{2} & 1 & 1 & 5 \\ 0 & \boxed{-8} & -2 & -12 \\ 0 & 0 & \boxed{1} & 2 \end{bmatrix}$$

Matrix-Vector and Matrix-Matrix multiplication

- ▶ Inner product of two vectors
- ▶ Ax : by rows, by columns
- ▶ AB : (i, j) entry is the inner product of i th row of A and j th column of B . Why do it this way? Composition of linear functions, $h = f \circ g$.
- ▶ Running time $O(n^2)$, Strassen $O(n^{\log_2 7})$.



- ▶ Associative, distributive, NOT commutative
- ▶ Matrix sums and scalar multiplication: done element-wise

EXAMPLE

$$\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 2 & 3 \\ 7 & 8 \end{bmatrix} = \begin{bmatrix} 7 & 8 \\ 2 & 3 \end{bmatrix} \text{ (permutation)}$$

LU Factorization

- ▶ **Elementary matrix** subtracts a multiple l of row j from row i : 1's on the diagonal, (i,j) entry is $-l$.

EXAMPLE

Gaussian elimination as GFE, applied to A,

$$GFE = \begin{bmatrix} 1 & & \\ & 1 & \\ & & 1 \end{bmatrix} \begin{bmatrix} 1 & & \\ & 1 & \\ & & 1 \end{bmatrix} \begin{bmatrix} 1 & & \\ -2 & 1 & \\ & & 1 \end{bmatrix} = \begin{bmatrix} 1 & & \\ -2 & 1 & \\ -1 & 1 & 1 \end{bmatrix}$$
$$\Rightarrow GFE \cdot A = U$$

“Reverse” Gaussian elimination as $E^{-1}F^{-1}G^{-1}$, applied to U,

$$E^{-1}F^{-1}G^{-1} = \begin{bmatrix} 1 & & \\ 2 & 1 & \\ & & 1 \end{bmatrix} \begin{bmatrix} 1 & & \\ & 1 & \\ -1 & & 1 \end{bmatrix} \begin{bmatrix} 1 & & \\ & 1 & \\ & -1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & & \\ 2 & 1 & \\ -1 & -1 & 1 \end{bmatrix},$$
$$L \equiv E^{-1}F^{-1}G^{-1} \Rightarrow \boxed{LU = A},$$

where U contains pivots and L contains multipliers from GE.

LU Factorization, comments

- ▶ A record of elimination steps, gives complete information.
- ▶ For multiple right hand sides, solve: $Lc = b$, then $Ux = c$, with $\frac{n^2}{2}$ operations each!
- ▶ How to write LU as LDU , where L and U have 1's on the diagonal and D is the diagonal matrix of pivots?

$$U = \begin{bmatrix} d_1 & & & \\ & d_2 & & \\ & & \ddots & \\ & & & d_n \end{bmatrix} \begin{bmatrix} 1 & u_{12}/d_1 & u_{12}/d_1 & \dots \\ & 1 & u_{23}/d_2 & \dots \\ & & \ddots & \vdots \\ & & & 1 \end{bmatrix}$$

- ▶ For a positive definite matrix (pivots positive), LDU reduces to RR^T (Cholesky factorization), where R is upper triangular.
- ▶ Not all non-singular matrices possess LU factors (leading principal sub-matrices must non-singular).

LU Factorization, examples

- ▶ With the rows reordered in advance, PA can be factored into LU , where P is a permutation matrix (how to find it?), e.g.

$$P_{13} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}, \quad P_{23} = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}.$$

EXAMPLE

Find LU factors of $A = \begin{bmatrix} 2 & 2 & 2 \\ 4 & 7 & 7 \\ 6 & 18 & 22 \end{bmatrix}$. To save computer memory, A can be successively overwritten with information in L and U , as GE evolves.

EXAMPLE

Use partial pivoting on $A = \begin{bmatrix} 1 & 2 & -3 & 4 \\ 4 & 8 & 12 & -8 \\ 2 & 3 & 2 & 1 \\ -3 & -1 & 1 & -4 \end{bmatrix}$ and determine the LU

decomposition $PA = LU$, where P is the associated permutation matrix. Hint: Adjoin a “permutation counter column” to keep track of row interchanges and remember to permute b as well, if available.

Inverses and Transposes

- ▶ A invertible, if $A^{-1}A = AA^{-1} = I$
- ▶ Simple for 2 by 2 matrix, or diagonal matrix
- ▶ $(AB)^{-1} = B^{-1}A^{-1}$, similarly for three or more matrices
- ▶ Why bother? $Ax = b \Rightarrow x = A^{-1}b$
- ▶ Calculation of A^{-1} : Gauss-Jordan elimination
Instead of stopping at U , we continue by subtracting multiples of a row from the rows above, until we reach I :

$$\boxed{[A|I] \rightarrow [U|L^{-1}] \rightarrow [I|A^{-1}]}$$

Not practical though (n^3 operations, sensitive to round-off errors).

- ▶ Multiple tests for invertibility: independent rows/columns, nonzero pivots/determinant/eigenvalues, $A^T A$ positive definite, full rank
- ▶ Transpose, $(AB)^T$, symmetric matrix

Ill-conditioned matrices and round-off errors

- ▶ Finite-precision mathematics, large numbers may swamp small numbers (different scales!).
- ▶ Ill-conditioned matrix is sensitive to small perturbations. Small pivots VS zero pivots. Condition number $\kappa = \|A\| \|A^{-1}\|$.

EXAMPLE

$$A = \begin{bmatrix} 1. & 1. \\ 1. & 1.0001 \end{bmatrix}. \text{ (ill-conditioned)}$$

$$\text{For } b = \begin{bmatrix} 2 \\ 2 \end{bmatrix}, x = \begin{bmatrix} 2 \\ 0 \end{bmatrix}, \text{ while for perturbed } b = \begin{bmatrix} 2 \\ 2.0001 \end{bmatrix}, x = \begin{bmatrix} 1 \\ 1 \end{bmatrix}!$$

EXAMPLE

$$A' = \begin{bmatrix} 0.0001 & 1. \\ 1. & 1. \end{bmatrix}. \text{ (well-conditioned)}$$

For $b = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, solve with roundoff to 3 places. Then try partial pivoting, i.e. exchange rows so as to maximize the magnitudes of pivots.