

# Matrices and Linear Systems

## Unit 2

# Linear Systems

$$a_{11}x_1 + a_{12}x_2 \cdots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 \cdots + a_{2n}x_n = b_2$$

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Overdetermined, underdetermined, square, homogeneous, number of solutions

# Matrices

- Column space  $\text{span}(A)$

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- Null space  $\text{ker}(A)$
- Rank  $\text{rank}(A)$
- Nullity  $\text{null}(A)$
- Nonsingular square matrix  $A \in \mathbf{R}^{n \times n}$

# Properties

$$A \in \mathbf{R}^{m \times p}, \quad B \in \mathbf{R}^{p \times n}, \quad C = AB \in \mathbf{R}^{n \times m}$$

$$c_{ij} = \sum_{k=1}^p a_{ik} b_{kj}$$

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Noncommutative ring

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Multiplication count

$$A \in \mathbf{R}^{m \times p}, \quad B \in \mathbf{R}^{p \times n}, \quad C \in \mathbf{R}^{n \times q}$$

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Coppersmith, Winograd, down to  $n^{2.376}$

Potential use of group theory, Robinson

# Block Multiplication

- Exercise 2.5
- If each product in the blocks can be formed, then block multiplication works

# Diagonally Dominant

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- Show the bound

## LU Factorization of Tridiagonal Matrices

$$\begin{pmatrix} d_1 & c_1 & 0 \\ a_2 & d_2 & c_2 \\ 0 & a_3 & d_3 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ l_2 & 1 & 0 \\ 0 & l_3 & 1 \end{pmatrix} \begin{pmatrix} u_1 & c_1 & 0 \\ 0 & u_2 & c_2 \\ 0 & 0 & u_3 \end{pmatrix}$$

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Show that  $|u_1| > |c_1|$ ,  $|u_2| > |c_2|$ ,  $|u_3| > 0$

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Exercise 2.10