



ICS141: Discrete Mathematics for Computer Science I

Dept. Information & Computer Sci., University of Hawaii

Jan Stelovsky

based on slides by Dr. Baek and Dr. Still

Originals by Dr. M. P. Frank and Dr. J.L. Gross

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Lecture 19

Chapter 3. The Fundamentals

3.8 Matrices

3.8 Matrices

- A ***matrix*** is a rectangular array of objects (usually numbers).
- An $m \times n$ (“ m by n ”) matrix has exactly m horizontal rows, and n vertical columns.

$$\begin{bmatrix} 2 & 3 \\ 5 & -1 \\ 7 & 0 \end{bmatrix}$$

A 3×2 matrix

- Plural of matrix = *matrices* (say MAY-trih-sees)
- An $n \times n$ matrix is called a *square* matrix



Applications of Matrices

- Tons of applications, including:
 - Solving systems of linear equations
 - Computer Graphics, Image Processing
 - Games
 - Models within many areas of Computational Science & Engineering
 - Quantum Mechanics, Quantum Computing
 - Many, many more...

Row and Column Order

- The rows in a matrix are usually indexed 1 to m from top to bottom.
- The columns are usually indexed 1 to n from left to right.
- Elements are indexed by row, then by column.

$$\mathbf{A} = [a_{ij}] = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}$$



Matrix Equality

- Two matrices **A** and **B** are considered equal iff they have the same number of rows, the same number of columns, and all their corresponding elements are equal.

$$\begin{bmatrix} 3 & 2 \\ -1 & 6 \end{bmatrix} \neq \begin{bmatrix} 3 & 2 & 0 \\ -1 & 6 & 0 \end{bmatrix}$$



Matrix Sums

- The *sum* $\mathbf{A} + \mathbf{B}$ of two matrices \mathbf{A} , \mathbf{B} (which **must** have the same number of rows, and the same number of columns) is the matrix (also with the same shape) given by adding corresponding elements of \mathbf{A} and \mathbf{B} .

$$\mathbf{A} + \mathbf{B} = [a_{ij} + b_{ij}]$$

$$\begin{bmatrix} 2 & 6 \\ 0 & -8 \end{bmatrix} + \begin{bmatrix} 9 & 3 \\ -11 & 3 \end{bmatrix} = \begin{bmatrix} 11 & 9 \\ -11 & -5 \end{bmatrix}$$

Matrix Products

- For an $m \times k$ matrix \mathbf{A} and a $k \times n$ matrix \mathbf{B} , the *product* \mathbf{AB} is the $m \times n$ matrix:

$$\mathbf{AB} = \mathbf{C} = [c_{ij}]$$

$$c_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{ik}b_{kj} = \sum_{\ell=1}^k a_{i\ell}b_{\ell j}$$

- I.e., the element of \mathbf{AB} indexed (i, j) is given by the vector *dot product* of the i -th row of \mathbf{A} and the j -th column of \mathbf{B} (considered as vectors).
- **Note:** Matrix multiplication is not commutative!

Matrix Product Example

$$\begin{bmatrix} 0 & 1 & -1 \\ 2 & 0 & 3 \end{bmatrix} \cdot \begin{bmatrix} 0 \\ 2 \\ 1 \end{bmatrix} \begin{bmatrix} -1 \\ 0 \\ 0 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{0 \cdot 0 + 1 \cdot 2 + (-1) \cdot 1}{2 \cdot 0 + 0 \cdot 2 + 3 \cdot 1} & \frac{0 \cdot (-1) + 1 \cdot 0 + (-1) \cdot 0}{2 \cdot (-1) + 0 \cdot 0 + 3 \cdot 0} \end{bmatrix}$$

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



Matrix Product Example

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 2 \\ 3 & 1 & 1 \end{bmatrix} \quad \text{and} \quad \mathbf{B} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

- Because **A** is a 2×3 matrix and **B** is a 2×2 matrix, the product **AB** is not defined.



Matrix Multiplication: Non-Commutative

- Matrix multiplication is not commutative!
- **A**: $m \times n$ matrix and **B**: $r \times s$ matrix
 - **AB** is defined when $n = r$
 - **BA** is defined when $s = m$
 - When both **AB** and **BA** are defined, generally they are not the same size unless $m = n = r = s$
 - If both **AB** and **BA** are defined and are the same size, then **A** and **B** must be square and of the same size
 - Even when **A** and **B** are both $n \times n$ matrices, **AB** and **BA** are not necessarily equal



Matrix Product Example

$$\mathbf{A} = \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix} \quad \text{and} \quad \mathbf{B} = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\mathbf{AB} = \begin{bmatrix} 3 & 2 \\ 5 & 3 \end{bmatrix} \quad \mathbf{BA} = \begin{bmatrix} 4 & 3 \\ 3 & 2 \end{bmatrix}$$

Matrix Multiplication Algorithm

procedure *matmul*(matrices **A**: $m \times k$, **B**: $k \times n$)

for $i := 1$ to m $\} \Theta(m) \cdot \{$
 for $j := 1$ to n begin $\} \Theta(n) \cdot ($
 $c_{ij} := 0$ $\} \Theta(1) +$
 for $q := 1$ to k $\} \Theta(k) \cdot$
 $c_{ij} := c_{ij} + a_{iq}b_{qj}$ $\} \Theta(1)) \}$

end $\{ \mathbf{C} = [c_{ij}]$ is the product of **A** and **B** $\}$

What's the Θ of its
time complexity?

Answer:
 $\Theta(mnk)$

Identity Matrices

- The **identity matrix** of order n , \mathbf{I}_n , is the rank- n square matrix with 1's along the upper-left to lower-right diagonal, and 0's everywhere else.

$$\mathbf{I}_n = [\delta_{ij}] = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

$\forall 1 \leq i, j \leq n$
 δ_{ij} is the Kronecker Delta
 n (rows)
 n (columns)



Matrix Inverses

- For some (but not all) square matrices \mathbf{A} , there exists a unique multiplicative ***inverse*** \mathbf{A}^{-1} of \mathbf{A} , a matrix such that $\mathbf{A}^{-1}\mathbf{A} = \mathbf{I}_n$.
- If the inverse exists, it is unique, and $\mathbf{A}^{-1}\mathbf{A} = \mathbf{A}\mathbf{A}^{-1}$.
- We won't go into the algorithms for matrix inversion...

Powers of Matrices

If \mathbf{A} is an $n \times n$ square matrix and $p \geq 0$, then:

- $\mathbf{A}^p = \underbrace{\mathbf{A}\mathbf{A}\mathbf{A}\cdots\mathbf{A}}_{p \text{ times}}$ (and $\mathbf{A}^0 = \mathbf{I}_n$)

- Example:
$$\begin{aligned} \begin{bmatrix} 2 & 1 \\ -1 & 0 \end{bmatrix}^3 &= \begin{bmatrix} 2 & 1 \\ -1 & 0 \end{bmatrix} \cdot \begin{bmatrix} 2 & 1 \\ -1 & 0 \end{bmatrix} \cdot \begin{bmatrix} 2 & 1 \\ -1 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 2 & 1 \\ -1 & 0 \end{bmatrix} \cdot \begin{bmatrix} 3 & 2 \\ -2 & -1 \end{bmatrix} \\ &= \begin{bmatrix} 4 & 3 \\ -3 & -2 \end{bmatrix} \end{aligned}$$

Matrix Transposition

- If $\mathbf{A} = [a_{ij}]$ is an $m \times n$ matrix, the **transpose** of \mathbf{A} (often written \mathbf{A}^t or \mathbf{A}^T) is the $n \times m$ matrix given by $\mathbf{A}^t = \mathbf{B} = [b_{ij}] = [a_{ji}]$ ($1 \leq i \leq n, 1 \leq j \leq m$)

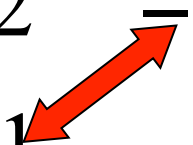
$$\begin{bmatrix} 2 & 1 & 3 \\ 0 & -1 & -2 \end{bmatrix}^t = \begin{bmatrix} 2 & 0 \\ 1 & -1 \\ 3 & -2 \end{bmatrix}$$

Symmetric Matrices

- A square matrix \mathbf{A} is ***symmetric*** iff $\mathbf{A} = \mathbf{A}^t$.
I.e., $\forall i, j \leq n: a_{ij} = a_{ji}$.
- Which of the below matrices is symmetric?

$$\begin{bmatrix} 1 & 1 \\ 1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$\begin{bmatrix} -2 & 1 & 3 \\ 1 & 0 & -1 \\ 3 & -1 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 3 & 0 & 1 \\ 0 & 2 & -1 \\ 1 & 1 & -2 \end{bmatrix}$$




Zero-One Matrices

- Useful for representing other structures.
 - *E.g.*, relations, directed graphs (later on)
- All elements of a *zero-one* matrix are either 0 or 1.
 - *E.g.*, representing **False** & **True** respectively.
- The *join* of **A**, **B** (both $m \times n$ zero-one matrices):
 - $\mathbf{A} \vee \mathbf{B} = [a_{ij} \vee b_{ij}]$
- The *meet* of **A**, **B**:
 - $\mathbf{A} \wedge \mathbf{B} = [a_{ij} \wedge b_{ij}] = [a_{ij} b_{ij}]$



Join and Meet Example

$$\mathbf{A} = \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \text{ and } \mathbf{B} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 1 & 0 \end{bmatrix}$$

$$\mathbf{A} \vee \mathbf{B} = \begin{bmatrix} 1 \vee 0 & 0 \vee 1 & 1 \vee 0 \\ 0 \vee 1 & 1 \vee 1 & 0 \vee 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

$$\mathbf{A} \wedge \mathbf{B} = \begin{bmatrix} 1 \wedge 0 & 0 \wedge 1 & 1 \wedge 0 \\ 0 \wedge 1 & 1 \wedge 1 & 0 \wedge 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$



Boolean Products

- Let $\mathbf{A} = [a_{ij}]$ be an $m \times k$ zero-one matrix and $\mathbf{B} = [b_{ij}]$ be a $k \times n$ zero-one matrix,
- The **boolean product** of \mathbf{A} and \mathbf{B} is like normal matrix multiplication, but using \vee instead of $+$, and \wedge instead of \times in the row-column “vector dot product”:

$$\mathbf{A} \odot \mathbf{B} = \mathbf{C} = [c_{ij}] = \left[\bigvee_{\ell=1}^k a_{i\ell} \wedge b_{\ell j} \right]$$

Boolean Products Example

- Find the Boolean product of **A** and **B**, where

$$\mathbf{A} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ 1 & 0 \end{bmatrix} \quad \mathbf{B} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \end{bmatrix}$$

$$\mathbf{A} \odot \mathbf{B} = \begin{bmatrix} (1 \wedge 1) \vee (0 \wedge 0) & (1 \wedge 1) \vee (0 \wedge 1) & (1 \wedge 0) \vee (0 \wedge 1) \\ (0 \wedge 1) \vee (1 \wedge 0) & (0 \wedge 1) \vee (1 \wedge 1) & (0 \wedge 0) \vee (1 \wedge 1) \\ (1 \wedge 1) \vee (0 \wedge 0) & (1 \wedge 1) \vee (0 \wedge 1) & (1 \wedge 0) \vee (0 \wedge 1) \end{bmatrix}$$

$$= \begin{bmatrix} 1 \vee 0 & 1 \vee 0 & 0 \vee 0 \\ 0 \vee 0 & 0 \vee 1 & 0 \vee 1 \\ 1 \vee 0 & 1 \vee 0 & 0 \vee 0 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ 1 & 1 & 0 \end{bmatrix}$$

Boolean Powers

- For a square zero-one matrix \mathbf{A} , and any $k \geq 0$, the *k-th Boolean power* of \mathbf{A} is simply the Boolean product of k copies of \mathbf{A} .

$$\mathbf{A}^{[k]} = \mathbf{A} \odot \mathbf{A} \odot \cdots \odot \mathbf{A}$$

$\underbrace{\hspace{10em}}_{k \text{ times}}$

$$\mathbf{A}^{[0]} = \mathbf{I}_n$$



Matrices as Functions

- An $m \times n$ matrix $\mathbf{A} = [a_{ij}]$ of members of a set S can be encoded as a partial function

$$f_{\mathbf{A}}: \mathbf{N} \times \mathbf{N} \rightarrow S,$$

such that for $i < m, j < n$, $f_{\mathbf{A}}(i, j) = a_{ij}$.

- By extending the domain over which $f_{\mathbf{A}}$ is defined, various types of infinite and/or multidimensional matrices can be obtained.