

Module #9: Matrices

Rosen 5th ed., §2.7
~18 slides, ~1 lecture

§2.7 Matrices



- A *matrix* (say MAY-trix) 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.
- Plural of matrix = *matrices* $\begin{bmatrix} 2 & 3 \\ 5 & -1 \\ 7 & 0 \end{bmatrix}$ a 3×2 matrix (say MAY-trih-sees)
- An $n \times n$ matrix is called a *square* matrix, whose *order* is n .

Not
our
meaning!

Note: The singular form of “matrices” is “*matrix*,” **not** “MAY-trih-see”!

Applications of Matrices

Tons of applications, including:

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

Matrix Equality

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

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

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 column.

$$\mathbf{A} = [a_{i,j}] = \begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,n} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m,1} & a_{m,2} & \cdots & a_{m,n} \end{bmatrix}$$

Matrices as Functions

- An $m \times n$ matrix $\mathbf{A} = [a_{i,j}]$ of members of a set S can be encoded as a partial function
$$f_{\mathbf{A}}: \mathbb{N} \times \mathbb{N} \rightarrow S,$$
such that for $i < m, j < n, f_{\mathbf{A}}(i, j) = a_{i,j}$.
- By extending the domain over which $f_{\mathbf{A}}$ is defined, various types of infinite and/or multidimensional matrices can be obtained.

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.

- $\mathbf{A}+\mathbf{B} = [a_{i,j}+b_{i,j}]$

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

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:

$$\mathbf{AB} = \mathbf{C} = [c_{i,j}] \equiv \left[\sum_{\ell=1}^k a_{i,\ell} b_{\ell,j} \right]$$

- *I.e.*, element (i,j) of \mathbf{AB} 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

- An example matrix multiplication to practice in class:

$$\begin{bmatrix} 0 & 1 & -1 \\ 2 & 0 & 3 \end{bmatrix} \cdot \begin{bmatrix} 0 & -1 & 1 & 0 \\ 2 & 0 & -2 & 0 \\ 1 & 0 & 3 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & -5 & -1 \\ 3 & -2 & 11 & 3 \end{bmatrix}$$

Identity Matrices

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

$$\mathbf{I}_n = \begin{bmatrix} \left\{ \begin{array}{l} 1 \text{ if } i = j \\ 0 \text{ if } i \neq j \end{array} \right. \\ \\ \\ \end{bmatrix} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

Review: §2.6 Matrices, so far

Matrix sums and products:

$$\mathbf{A} + \mathbf{B} = [a_{i,j} + b_{i,j}]$$

$$\mathbf{AB} = \mathbf{C} = [c_{i,j}] \equiv \left[\sum_{\ell=1}^k a_{i,\ell} b_{\ell,j} \right]$$

Identity matrix of order n :

$$\mathbf{I}_n = [\delta_{ij}], \text{ where } \delta_{ij} = 1 \text{ if } i=j \text{ and } \delta_{ij} = 0 \text{ if } i \neq j.$$

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...

Matrix Multiplication Algorithm

```

procedure matmul(matrices A:  $m \times k$ , B:  $k \times n$ )
for  $i := 1$  to  $m$ 
    for  $j := 1$  to  $n$  begin
         $c_{ij} := 0$ 
        for  $q := 1$  to  $k$ 
             $c_{ij} := c_{ij} + a_{iq}b_{qj}$ 
        end {C=[ $c_{ij}$ ] is the product of A and B}
    end
end
    
```

$\Theta(m) \cdot$
 $\Theta(n) \cdot$
 $\Theta(1) +$
 $\Theta(k) \cdot$
 $\Theta(1)$

What's the Θ of its time complexity?

Answer:
 $\Theta(mnk)$

Powers of Matrices

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

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

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

- Example:

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

$$= \begin{bmatrix} 4 & 3 \\ -3 & -2 \end{bmatrix}$$

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 $\mathbf{A}^t = \mathbf{B} = [b_{ij}] = [a_{ji}]$ ($1 \leq i \leq n, 1 \leq j \leq m$)

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

Flip
across
diagonal

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 is symmetric?

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

Zero-One Matrices

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

Boolean Products

- Let $\mathbf{A}=[a_{ij}]$ be an $m \times k$ zero-one matrix, & let $\mathbf{B}=[b_{ij}]$ be a $k \times n$ zero-one matrix,
- The *boolean product* of \mathbf{A} and \mathbf{B} is like normal matrix \times , but using \vee instead $+$ 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 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} .

- $\mathbf{A}^{[k]} \equiv \underbrace{\mathbf{A} \odot \mathbf{A} \odot \dots \odot \mathbf{A}}_{k \text{ times}}$