

# Lecture 6 - Matrix Operations

Note Title

2/4/2008

Today we're starting a slightly different look at systems et al.

Terms form matrices:

Size = # rows x # columns

Elements are indexed by a pair of numbers:  $a_{i,j}$  is the entry in row  $i$ , column  $j$

main diagonal = entries with  $i=j$ :  $a_{1,1}, a_{2,2}, \dots$

$A$  &  $B$  are equal if they are the same size and  $a_{i,j} = b_{i,j} \quad \forall i, j$

If  $A$  &  $B$  are the same size, then the sum  $C = A + B$  is the matrix with

$$c_{i,j} = a_{i,j} + b_{i,j}.$$

Can also scale a matrix by scaling every entry:

$$B = cA \leftrightarrow b_{i,j} = ca_{i,j}$$

Ex:  $A = \begin{bmatrix} 0 & 1 \\ 1 & 2 \\ 3 & 5 \end{bmatrix}$ ,  $B = \begin{bmatrix} 1 & 1 \\ 2 & 3 \\ 5 & 8 \end{bmatrix}$ ,  $C = \begin{bmatrix} 13 & 21 \\ 34 & 55 \end{bmatrix}$   $\Rightarrow$   $A$  is  $3 \times 2$ ,  $B$  is  $3 \times 2$ ,  $C$  is  $2 \times 2$

— = main diagonal.

Then  $a_{3,2} = 5 = b_{3,1}$ ,  $A+B = \begin{bmatrix} 1 & 2 \\ 3 & 5 \\ 8 & 13 \end{bmatrix}$ ,  $A+C$  not defined.

$$2A = \begin{bmatrix} 0 & 2 \\ 2 & 4 \\ 6 & 10 \end{bmatrix} \quad 2C = \begin{bmatrix} 26 & 42 \\ 68 & 110 \end{bmatrix}, \quad 0 \cdot C = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

For every size, there is a zero matrix  $O_{m,n}$  in which all entries are zero.

Properties:

1)  $(A+B)+C = A+(B+C)$

2)  $A+O = O+A = A$

3)  $A+(-1)A = (-1)A+A = O$

4)  $A+B = B+A$

5)  $1 \cdot A = A$

6)  $a \cdot (b \cdot c) = (a \cdot b) \cdot c$

7)  $a \cdot (B+C) = a \cdot B + a \cdot C$

8)  $(a+b) \cdot c = a \cdot c + b \cdot c$

This shows that  $m \times n$  matrices forms a vector space (basically  $\mathbb{R}^{mn}$ ).

Real power comes from matrix multiplication.

If  $A$  is  $m \times n$  and  $B$  is  $n \times k$ , then  $AB$  is defined and is  $m \times k$ .

Fast def: The rows of  $A$  are vectors in  $\mathbb{R}^n$   
 $A = \begin{bmatrix} \bar{a}^1 \\ \vdots \\ \bar{a}^m \end{bmatrix}$        $\bar{a}^i$  are row vectors

The columns of  $B$  are also vectors in  $\mathbb{R}^n$   
 $B = \begin{bmatrix} \bar{b}_1 & \dots & \bar{b}_k \end{bmatrix}$

The  $C = AB$  has  $C_{ij} = \bar{a}^i \cdot \bar{b}_j$ .

Ex: 1)  $\begin{bmatrix} 1 & 2 \\ 0 & 3 \end{bmatrix}_A \cdot \begin{bmatrix} 1 & 1 & 3 \\ 5 & 7 & 9 \end{bmatrix}_B = \begin{bmatrix} 1 \cdot 1 + 2 \cdot 5 & 1 \cdot 1 + 2 \cdot 7 & 1 \cdot 3 + 2 \cdot 9 \\ 0 \cdot 1 + 3 \cdot 5 & 0 \cdot 1 + 3 \cdot 7 & 0 \cdot 3 + 3 \cdot 9 \end{bmatrix}$

2)  $\begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix}_A \cdot \begin{bmatrix} a \\ b \\ c \end{bmatrix}_B = \begin{bmatrix} a + 2b + 3c \\ 4a + 5b + 6c \\ 7a + 8b + 9c \end{bmatrix}$

Goal: rewrite systems as  $A\bar{x} = \bar{b}$ , where  $\bar{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ ,  $\bar{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$ ,  $A_{m \times n}$ .

$\leftrightarrow$  multiply  $A$   $m \times n$  by  $\bar{x}$   $n \times 1$ , getting  $\bar{b}$   $m \times 1$  and  
 $A = [a_{ij}]$ , then  $A\bar{x} = \begin{bmatrix} a_{11}x_1 + \dots + a_{1n}x_n \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$

$\leftrightarrow \left. \begin{matrix} a_{11}x_1 + \dots + a_{1n}x_n = b_1 \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n = b_m \end{matrix} \right\} \text{number of eqns} = \# \text{ of rows in } A$

# of var =  
 # of col of  
 $A$

Def A diagonal matrix is a square matrix such that the non-diagonal elements are 0.

An identity matrix is a diagonal matrix with ones on the diagonal.

Properties of Matrix mult:

- 1)  $A(BC) = (AB)C$  - associativity
  - 2)  $A \text{ } m \times n \Rightarrow I_m \cdot A = A \cdot I_n = A$
  - 3)  $A(B+C) = AB+AC$
  - 4)  $(A+B)C = AC+BC$
  - 5)  $A \cdot cB = cA \cdot B = c(A \cdot B)$  - matrix and scalar mults commute
- } matrix mult distributes over +

Application: The solutions to a homogeneous system form a subspace.

Why? Let  $\bar{x}$  and  $\bar{y}$  be solutions to  $A\bar{x} = \bar{0}$ . Then we need to show

$\bar{x} + \bar{y}$  and  $k\bar{x}$  are also solutions:

$$A \cdot (\bar{x} + \bar{y}) = A\bar{x} + A\bar{y} = \bar{0} + \bar{0} = \bar{0} \quad \checkmark$$

$$A \cdot (k\bar{x}) = k(A\bar{x}) = k \cdot \bar{0} = \bar{0} \quad \checkmark$$

Could have done this directly. This was a lot easier.

One last topic: Block matrices.

Def A block partition of a matrix is a grouping of rows and columns to form submatrices.

Ex: 
$$\left[ \begin{array}{c|cc} 1 & 2 & 3 \\ \hline 3 & 9 & 14 \\ \hline 0 & 1 & 6 \end{array} \right] \rightsquigarrow \begin{bmatrix} A & B \\ C & D \end{bmatrix} \text{ where } A = \begin{bmatrix} 1 \\ 3 \end{bmatrix}, B = \begin{bmatrix} 2 & 3 \\ 9 & 14 \end{bmatrix}, C = [0], D = [1 \ 6]$$

are matrices.

Important simplification: Block matrices block multiply:

$$A = \begin{bmatrix} P & Q \\ R & S \end{bmatrix}, B = \begin{bmatrix} M & N \\ U & V \end{bmatrix}, AB = \begin{bmatrix} PM+QU & PN+QV \\ RM+SU & RN+SV \end{bmatrix} \quad (\text{when they all make sense})$$

For this to work: # columns of  $P \neq R =$  # Rows  $M \neq N$

# columns of  $Q \neq S =$  # Rows of  $U \neq V$

So the column partition of A must agree w/ the row partition of B.