

A First Course in  
**LINEAR ALGEBRA**

**Lecture Notes**  
for Math 1503

**Matrices: Matrix Arithmetic**

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A First Course in Linear Algebra

Lecture Slides

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# Matrices - Basic Definitions and Notation

## Definitions

Let  $m$  and  $n$  be positive integers.

- An  $m \times n$  matrix is a rectangular array of numbers having  $m$  rows and  $n$  columns. Such a matrix is said to have size  $m \times n$ .
- A row matrix (or row) is a  $1 \times n$  matrix, and a column matrix (or column) is an  $m \times 1$  matrix.
- A square matrix is an  $n \times n$  matrix.
- The  $(i, j)$ -entry of a matrix is the entry in row  $i$  and column  $j$ . For a matrix  $A$ , the  $(i, j)$ -entry of  $A$  is often written as  $a_{ij}$ .

General notation for an  $m \times n$  matrix,  $A$ :

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \cdots & a_{mn} \end{bmatrix} = [a_{ij}]$$

## Matrices – Properties and Operations

- 1 **Equality:** two matrices are equal if and only if they have the same size and the corresponding entries are equal.
- 2 **Zero Matrix:** an  $m \times n$  matrix with all entries equal to zero.
- 3 **Addition:** matrices must have the same size; add corresponding entries.
- 4 **Scalar Multiplication:** multiply each entry of the matrix by the scalar.
- 5 **Negative of a Matrix:** for an  $m \times n$  matrix  $A$ , its negative is denoted  $-A$  and  $-A = (-1)A$ .
- 6 **Subtraction:** for  $m \times n$  matrices  $A$  and  $B$ ,  $A - B = A + (-1)B$ .

# Matrix Addition

## Definition

Let  $A = [a_{ij}]$  and  $B = [b_{ij}]$  be two  $m \times n$  matrices. Then  $A + B = C$  where  $C$  is the  $m \times n$  matrix  $C = [c_{ij}]$  defined by

$$c_{ij} = a_{ij} + b_{ij}$$

## Example

Let  $A = \begin{bmatrix} 1 & 3 \\ 2 & 5 \end{bmatrix}$ ,  $B = \begin{bmatrix} 0 & -2 \\ 6 & 1 \end{bmatrix}$ . Then,

$$A + B =$$

## Theorem (Properties of Matrix Addition)

Let  $A, B$  and  $C$  be  $m \times n$  matrices. Then the following properties hold.

- ①  $A + B = B + A$  (matrix addition is commutative).
- ②  $(A + B) + C = A + (B + C)$  (matrix addition is associative).
- ③ There exists an  $m \times n$  zero matrix,  $0$ , such that  $A + 0 = A$ .  
(existence of an additive identity).
- ④ There exists an  $m \times n$  matrix  $-A$  such that  $A + (-A) = 0$ .  
(existence of an additive inverse).

# Scalar Multiplication

## Definition

Let  $A = [a_{ij}]$  be an  $m \times n$  matrix and let  $k$  be a scalar. Then  $kA = [ka_{ij}]$ .

## Example

Let  $A = \begin{bmatrix} 2 & 0 & -1 \\ 3 & 1 & -2 \\ 0 & 4 & 5 \end{bmatrix}$ .

Then

$$3A =$$

## Theorem (Properties of Scalar Multiplication)

Let  $A, B$  be  $m \times n$  matrices and let  $k, p \in \mathbb{R}$  (scalars). Then the following properties hold.

- ①  $k(A + B) = kA + kB$ .  
(scalar multiplication distributes over matrix addition).
- ②  $(k + p)A = kA + pA$ .  
(addition distributes over scalar multiplication).
- ③  $k(pA) = (kp)A$ . (scalar multiplication is associative).
- ④  $1A = A$ . (existence of a multiplicative identity).

### Example

$$2 \begin{bmatrix} -1 & 0 \\ 1 & 1 \end{bmatrix} + 4 \begin{bmatrix} -2 & 1 \\ 3 & 0 \end{bmatrix} - \begin{bmatrix} 6 & 8 \\ 1 & -1 \end{bmatrix} =$$

### Problem

Let  $A$  and  $B$  be  $m \times n$  matrices. Simplify the expression

$$2[9(A - B) + 7(2B - A)] - 2[3(2B + A) - 2(A + 3B) - 5(A + B)]$$

### Solution

$$\begin{aligned} & 2[9(A - B) + 7(2B - A)] - 2[3(2B + A) - 2(A + 3B) - 5(A + B)] \\ = & 2(9A - 9B + 14B - 7A) - 2(6B + 3A - 2A - 6B - 5A - 5B) \\ = & 2(2A + 5B) - 2(-4A - 5B) \\ = & 12A + 20B \end{aligned}$$

## Vectors

### Definitions

A row matrix or column matrix is often called a **vector**, and such matrices are referred to as **row vectors** and **column vectors**, respectively. If  $X$  is a row vector of size  $1 \times n$ , and  $Y$  is a column vector of size  $m \times 1$ , then we write

$$X = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \text{ and } Y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$

## Vector form of a system of linear equations

### Definition

Consider the system of linear equations

$$\begin{array}{ccccccc} a_{11}x_1 & + & a_{12}x_2 & + & \cdots & + & a_{1n}x_n & = & b_1 \\ a_{21}x_1 & + & a_{22}x_2 & + & \cdots & + & a_{2n}x_n & = & b_2 \\ \vdots & & \vdots & & & & \vdots & & \vdots \\ a_{m1}x_1 & + & a_{m2}x_2 & + & \cdots & + & a_{mn}x_n & = & b_m \end{array}$$

Such a system can be expressed in **vector form** or as a **vector equation** by using **linear combinations** of column vectors:

$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

## Vector form of a system of linear equations

### Problem

Express the following system of linear equations in vector form.

$$\begin{array}{ccccccc} 2x_1 & + & 4x_2 & - & 3x_3 & = & -6 \\ & & - & x_2 & + & 5x_3 & = & 0 \\ x_1 & + & x_2 & + & 4x_3 & = & 1 \end{array}$$

### Solution

$$x_1 \begin{bmatrix} 2 \\ 0 \\ 1 \end{bmatrix} + x_2 \begin{bmatrix} 4 \\ -1 \\ 1 \end{bmatrix} + x_3 \begin{bmatrix} -3 \\ 5 \\ 4 \end{bmatrix} = \begin{bmatrix} -6 \\ 0 \\ 1 \end{bmatrix}$$

## Matrix Vector Multiplication

### Definition

Let  $A = [a_{ij}]$  be an  $m \times n$  matrix with columns  $A_1, A_2, \dots, A_n$ , written  $A = [A_1 \ A_2 \ \cdots \ A_n]$ , and let  $X$  be an  $n \times 1$  column vector,

$$X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

Then the product of matrix  $A$  and (column) vector  $X$  is the  $m \times 1$  column vector given by

$$\begin{bmatrix} A_1 & A_2 & \cdots & A_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1 A_1 + x_2 A_2 + \cdots + x_n A_n = \sum_{j=1}^n x_j A_j$$

that is,  $AX$  is a linear combination of the columns of  $A$ . Notice how this is a generalization of the dot product between vectors.

## Matrix Vector Multiplication

### Problem

Compute the product  $AX$  for

$$A = \begin{bmatrix} 1 & 4 \\ 5 & 0 \end{bmatrix} \text{ and } X = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$

### Solution

$$AX = \begin{bmatrix} 1 & 4 \\ 5 & 0 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \end{bmatrix} = 2 \begin{bmatrix} 1 \\ 5 \end{bmatrix} + 3 \begin{bmatrix} 4 \\ 0 \end{bmatrix} = \begin{bmatrix} 2 \\ 10 \end{bmatrix} + \begin{bmatrix} 12 \\ 0 \end{bmatrix} = \begin{bmatrix} 14 \\ 10 \end{bmatrix}$$

# Matrix Vector Multiplication

## Problem

Compute  $AY$  for

$$A = \begin{bmatrix} 1 & 0 & 2 & -1 \\ 2 & -1 & 0 & 1 \\ 3 & 1 & 3 & 1 \end{bmatrix} \text{ and } Y = \begin{bmatrix} 2 \\ -1 \\ 1 \\ 4 \end{bmatrix}$$

## Solution

$AY =$

# Matrix form of a system of linear equations

## Definition

Consider the system of linear equations

$$\begin{array}{ccccccc} a_{11}x_1 & + & a_{12}x_2 & + & \cdots & + & a_{1n}x_n & = & b_1 \\ a_{21}x_1 & + & a_{22}x_2 & + & \cdots & + & a_{2n}x_n & = & b_2 \\ \vdots & & \vdots & & & & \vdots & & \vdots \\ a_{m1}x_1 & + & a_{m2}x_2 & + & \cdots & + & a_{mn}x_n & = & b_m \end{array}$$

Such a system can be expressed in **matrix form** using matrix vector multiplication,

$$\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Thus a system of linear equations can be expressed as a **matrix equation**  $AX = B$ , where  $A$  is the coefficient matrix,  $B$  is the constant matrix, and  $X$  is the matrix of variables.



## Matrix form of a system of linear equations

### Problem

Express the following system of linear equations in matrix form.

$$\begin{array}{rrcrcl} 2x_1 & + & 4x_2 & - & 3x_3 & = & -6 \\ & & - & x_2 & + & 5x_3 & = & 0 \\ x_1 & + & x_2 & + & 4x_3 & = & 1 \end{array}$$

### Solution

$$\begin{bmatrix} 2 & 4 & -3 \\ 0 & -1 & 5 \\ 1 & 1 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -6 \\ 0 \\ 1 \end{bmatrix}$$

## Matrix and Vector Equations

### Theorem

- 1 Every system of  $m$  linear equations in  $n$  variables can be written in the form  $AX = B$  where  $A$  is the coefficient matrix,  $X$  is the matrix of variables, and  $B$  is the constant matrix.
- 2 The system  $AX = B$  is consistent (i.e., has at least one solution) if and only if  $B$  is a linear combination of the columns of  $A$ .

- 3 The vector  $X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$  is a solution to the system  $AX = B$  if and only if  $x_1, x_2, \dots, x_n$  are a solution to the vector equation

$$x_1 A_1 + x_2 A_2 + \cdots x_n A_n = B$$

where  $A_1, A_2, \dots, A_n$  are the columns of  $A$ .

## Proof of the Theorem (a sketch)

Every statement that deserves to be called a theorem deserves a proof, and the theorem from the previous slide is no exception. In this particular case the proof is straightforward (i.e. uneventful).

### Proof.

(a) One first checks that  $(x_1, \dots, x_n)$  is a solution to the original system if and only if  $X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$  is a solution to  $AX = B$ .

This depends on the way that the matrix arithmetics (addition, multiplication by scalars, multiplication) was defined.

## Proof continued

### Proof.

(b) Once (a) is taken care of, it gives a one-to-one correspondence between the set of solutions to the original system and the set of solutions to  $AX = B$ :

$$(x_1, \dots, x_n) \mapsto \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}.$$

This is (3), and it implies that the two sets have the same cardinality, and (2) follows. □

## Problem

Let

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

Express  $B$  as a linear combination of the columns  $A_1, A_2, A_3, A_4$  of  $A$ , or show that this is impossible.

## Solution

Solve the system  $AX = B$  where  $X$  is a column vector with four entries. Do so by putting the **augmented matrix**  $[A \mid B]$  in reduced row-echelon form.

$$\left[ \begin{array}{cccc|c} 1 & 0 & 2 & -1 & 1 \\ 2 & -1 & 0 & 1 & 1 \\ 3 & 1 & 3 & 1 & 1 \end{array} \right] \rightarrow \dots \rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & 0 & 1 & \frac{1}{7} \\ 0 & 1 & 0 & 1 & -\frac{5}{7} \\ 0 & 0 & 1 & -1 & \frac{3}{7} \end{array} \right]$$

Since there are infinitely many solutions ( $x_4$  is assigned a parameter), choose any value for  $x_4$ . Choosing  $x_4 = 0$  (which is the simplest thing to do) gives us

$$B = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{7} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} - \frac{5}{7} \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} + \frac{3}{7} \begin{bmatrix} 2 \\ 0 \\ 3 \end{bmatrix} = \frac{1}{7}A_1 - \frac{5}{7}A_2 + \frac{3}{7}A_3 + 0A_4.$$

## Matrix Multiplication

### Definition (Product of two matrices)

Let  $A$  be an  $m \times n$  matrix and let  $B = [B_1 \ B_2 \ \dots \ B_p]$  be an  $n \times p$  matrix, whose columns are  $B_1, B_2, \dots, B_p$ . The **product of  $A$  and  $B$**  is the matrix

$$AB = A [B_1 \ B_2 \ \dots \ B_p] = [AB_1 \ AB_2 \ \dots \ AB_p]$$

i.e., the first column of  $AB$  is  $AB_1$ , the second column of  $AB$  is  $AB_2$ , etc. Note that  $AB$  has size  $m \times p$ .

### Definition (The $(i, j)$ -entry of a product)

Let  $A = [a_{ij}]$  be an  $m \times n$  matrix and  $B = [b_{ij}]$  be an  $n \times p$  matrix. Then the  $(i, j)$ -entry of  $AB$  is given by

$$a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj} = \sum_{k=1}^n a_{ik}b_{kj}$$

(Note: This can simply be viewed as the dot product of the  $i$ 'th row of  $A$  with the  $j$ 'th column of  $B$ .)

### Example

Using the above definition, the  $(2, 3)$ -entry of the product

$$\begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \begin{bmatrix} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{bmatrix}$$

is computed using the **second row** of the first matrix, and the **third column** of the second matrix, resulting in

$$2(2) + (-1)(4) + 1(0) = 4 - 4 + 0 = 0.$$

### Problem

Find the product  $AB$  of matrices

$$A = \begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix} \text{ and } B = \begin{bmatrix} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{bmatrix}$$

### Solution

# Compatibility for Matrix Multiplication

## Definition

Let  $A$  and  $B$  be matrices, and suppose that  $A$  is  $m \times n$ .

- In order for the product  $AB$  to exist, the number of rows in  $B$  must be equal to the number of columns in  $A$ , implying that  $B$  is an  $n \times p$  matrix for some  $p$ .
- When defined,  $AB$  is an  $m \times p$  matrix.

If the product is defined, then  $A$  and  $B$  are said to be **compatible** for (matrix) multiplication.

## Example

As we saw in the previous problem

$$\begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix}^{2 \times 3} \begin{bmatrix} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{bmatrix}^{3 \times 3} = \begin{bmatrix} 4 & -1 & -2 \\ -1 & 4 & 0 \end{bmatrix}^{2 \times 3}$$

Note that the product

$$\begin{bmatrix} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{bmatrix}^{3 \times 3} \begin{bmatrix} -1 & 0 & 3 \\ 2 & -1 & 1 \end{bmatrix}^{2 \times 3}$$

does not exist.

## Multiplication by the Zero Matrix

### Example

Compute the product  $A0$  for the matrix

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

and the  $2 \times 2$  zero matrix given by  $0 = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$

### Solution

*In this product, we compute*

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

*Hence,  $A0 = 0$ .*

## Questions on Matrix Multiplication

Given matrices  $A$  and  $B$ , is  $AB = BA$ ?

Suppose  $A$  is an  $m \times n$  matrix and  $B$  is an  $m' \times n'$  matrix.

The product  $AB$  is defined if and only if  $n = m'$ .

The product  $BA$  is defined if and only if  $m = n'$ .

Therefore the equation  $AB = BA$  makes sense if and only if  $A$  is an  $m \times n$  matrix and  $B$  is an  $n \times m$  matrix for some—possibly different— $m$  and  $n$ .

So the right question is:

Given matrices  $A$  and  $B$  such that both  $AB$  and  $BA$  are defined, is  $AB = BA$ ?

# Matrix Multiplication is Not Commutative

## Problem

Let

$$A = \begin{bmatrix} 1 & 2 \\ -3 & 0 \\ 1 & -4 \end{bmatrix} \text{ and } B = \begin{bmatrix} 1 & -1 & 2 & 0 \\ 3 & -2 & 1 & -3 \end{bmatrix}$$

- Does  $AB$  exist? If so, compute it.
- Does  $BA$  exist? If so, compute it.

## Solution

$$AB = \begin{bmatrix} 7 & -5 & 4 & -6 \\ -3 & 3 & -6 & 0 \\ -11 & 7 & -2 & 12 \end{bmatrix}$$

$BA$  does not exist

## Problem

Let

$$G = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \text{ and } H = \begin{bmatrix} 1 & 0 \end{bmatrix}$$

- Does  $GH$  exist? If so, compute it.
- Does  $HG$  exist? If so, compute it.

## Solution

In this example,  $GH$  and  $HG$  both exist, but they are not equal. They aren't even the same size!

## Problem

Let

$$P = \begin{bmatrix} 1 & 0 \\ 2 & -1 \end{bmatrix} \text{ and } Q = \begin{bmatrix} -1 & 1 \\ 0 & 3 \end{bmatrix}$$

- Does  $PQ$  exist? If so, compute it.
- Does  $QP$  exist? If so, compute it.

## Solution

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

$$QP = \begin{bmatrix} 1 & -1 \\ 6 & -3 \end{bmatrix}$$

In this example,  $PQ$  and  $QP$  both exist and are the same size, but  $PQ \neq QP$ .

## Fact

The three preceding problems illustrate an important property of matrix multiplication.

*In general, matrix multiplication is **not** commutative, i.e., the order of the matrices in the product is important.*

In other words, in general  $AB \neq BA$ .



## Problem

Let

$$U = \begin{bmatrix} 2 & 0 \\ 0 & 2 \end{bmatrix} \text{ and } V = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

- Does  $UV$  exist? If so, compute it.
- Does  $VU$  exist? If so, compute it.

## Solution

$$UV = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

$$VU = \begin{bmatrix} 2 & 4 \\ 6 & 8 \end{bmatrix}$$

In this particular example, the matrices **commute**, i.e.,  $UV = VU$ .

## Properties of Matrix Multiplication

### Theorem

Let  $A$ ,  $B$ , and  $C$  be matrices of the appropriate sizes, and let  $r \in \mathbb{R}$  be a scalar. Then the following properties hold.

- 1  $A(B + C) = AB + AC$ .  
(matrix multiplication distributes over matrix addition).
- 2  $(B + C)A = BA + CA$ .  
(matrix multiplication distributes over matrix addition).
- 3  $A(BC) = (AB)C$ . (matrix multiplication is associative).
- 4  $r(AB) = (rA)B = A(rB)$ .

This applies to matrix-vector multiplication as well, since a vector is a row matrix or a column matrix.

## Elementary Proofs

### Problem

Let  $A$  and  $B$  be  $m \times n$  matrices, and let  $C$  be an  $n \times p$  matrix. Prove that if  $A$  and  $B$  commute with  $C$ , then  $A + B$  commutes with  $C$ .

### Proof.



### Problem

Let  $A$ ,  $B$  and  $C$  be  $n \times n$  matrices, and suppose that both  $A$  and  $B$  commute with  $C$ , i.e.,  $AC = CA$  and  $BC = CB$ . Show that  $AB$  commutes with  $C$ .

### Proof.

We must show that  $(AB)C = C(AB)$  given that  $AC = CA$  and  $BC = CB$ .

$$\begin{aligned}(AB)C &= A(BC) \text{ (matrix multiplication is associative)} \\ &= A(CB) \text{ (} B \text{ commutes with } C \text{)} \\ &= (AC)B \text{ (matrix multiplication is associative)} \\ &= (CA)B \text{ (} A \text{ commutes with } C \text{)} \\ &= C(AB) \text{ (matrix multiplication is associative)}\end{aligned}$$

Therefore,  $AB$  commutes with  $C$ .



## Definition (Matrix Transpose)

If  $A$  is an  $m \times n$  matrix, then its **transpose**, denoted  $A^T$ , is the  $n \times m$  whose  $i^{\text{th}}$  row is the  $i^{\text{th}}$  column of  $A$ ,  $1 \leq i \leq n$ ; i.e., if  $A = [a_{ij}]$ , then

$$A^T = [a_{ij}]^T = [a_{ji}]$$

i.e., the  $(i, j)$ -entry of  $A^T$  is the  $(j, i)$ -entry of  $A$ .

## Theorem (Properties of the Transpose of a Matrix)

Let  $A$  and  $B$  be  $m \times n$  matrices,  $C$  be a  $n \times p$  matrix, and  $r \in \mathbb{R}$  a scalar. Then

①  $(A^T)^T = A$

③  $(A + B)^T = A^T + B^T$

②  $(rA)^T = rA^T$

④  $(AC)^T = C^T A^T$

To prove each these properties, you only need to compute the  $(i, j)$ -entries of the matrices on the left-hand side and the right-hand side. **And you can do it!**

## Problem

Find the matrix  $A$  if  $\left(A + 3 \begin{bmatrix} 1 & -1 & 0 \\ 1 & 2 & 4 \end{bmatrix}\right)^T = \begin{bmatrix} 2 & 1 \\ 0 & 5 \\ 3 & 8 \end{bmatrix}$ .

## Solution

# Symmetric Matrices

## Definition

Let  $A = [a_{ij}]$  be an  $m \times n$  matrix. The entries  $a_{11}, a_{22}, a_{33}, \dots$  are called the **main diagonal** of  $A$ .

## Definition

The matrix  $A$  is called **symmetric** if and only if  $A^T = A$ . Note that this immediately implies that  $A$  is a **square** matrix.

## Examples

$$\begin{bmatrix} 2 & -3 \\ -3 & 17 \end{bmatrix}, \begin{bmatrix} -1 & 0 & 5 \\ 0 & 2 & 11 \\ 5 & 11 & -3 \end{bmatrix}, \begin{bmatrix} 0 & 2 & 5 & -1 \\ 2 & 1 & -3 & 0 \\ 5 & -3 & 2 & -7 \\ -1 & 0 & -7 & 4 \end{bmatrix}$$

are symmetric matrices, and each is symmetric about its main diagonal.

## Problem

Show that if  $A$  and  $B$  are symmetric matrices, then  $2A - B$  is symmetric.

## Proof.



## Skew Symmetric Matrices

### Definition

An  $n \times n$  matrix  $A$  is said to be **skew symmetric** if  $A^T = -A$ .

### Example (Skew Symmetric Matrices)

$$\begin{bmatrix} 0 & 2 \\ -2 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 9 & 4 \\ -9 & 0 & -3 \\ -4 & 3 & 0 \end{bmatrix}$$

### Problem

Show that if  $A$  is a square matrix, then  $A - A^T$  is skew-symmetric.

### Solution

We must show that  $(A - A^T)^T = -(A - A^T)$ . Using the properties of matrix addition, scalar multiplication, and transposition

$$(A - A^T)^T = A^T - (A^T)^T = A^T - A = -(A - A^T).$$

## The $n \times n$ Identity Matrix

### Definition

For each  $n \geq 2$ , the  **$n \times n$  identity matrix**, denoted  $I_n$ , is the matrix having ones on its main diagonal and zeros elsewhere, and is defined for all  $n \geq 2$ .

### Example

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

### Definition

Let  $n \geq 2$ . For each  $j$ ,  $1 \leq j \leq n$ , we denote by  $E_j$  the  $j^{\text{th}}$  column of  $I_n$ .

### Example

$$\text{When } n = 3, E_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, E_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, E_3 = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}.$$

## Theorem

Let  $A$  be an  $m \times n$  matrix. Then  $AI_n = A$  and  $I_m A = A$ .

## Proof

The  $(i, j)$ -entry of  $AI_n$  is the product of the  $i^{\text{th}}$  row of  $A = [a_{ij}]$ , namely  $[a_{i1} \ a_{i2} \ \cdots \ a_{ij} \ \cdots \ a_{in}]$  with the  $j^{\text{th}}$  column of  $I_n$ , namely  $E_j$ . Since  $E_j$  has a one in row  $j$  and zeros elsewhere,

$$[a_{i1} \ a_{i2} \ \cdots \ a_{ij} \ \cdots \ a_{in}] E_j = a_{ij}$$

Since this is true for all  $i \leq m$  and all  $j \leq n$ ,  $AI_n = A$ .

The proof of  $I_m A = A$  is analogous—work it out!

Instead of  $AI_n$  and  $I_m A$  we often write  $AI$  and  $IA$ , respectively, since the size of the identity matrix is clear from the context: the sizes of  $A$  and  $I$  must be compatible for matrix multiplication.

Thus

$$AI = A \text{ and } IA = A$$

which is why  $I$  is called an **identity** matrix – it is an identity for matrix multiplication.

# Matrix Inverses

## Definition

Let  $A$  be an  $n \times n$  matrix. Then  $B$  is an **inverse** of  $A$  if and only if  $AB = I_n$  and  $BA = I_n$ . Note that since  $A$  and  $I_n$  are both  $n \times n$ ,  $B$  **must also be** an  $n \times n$  matrix.

## Example

Let  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$  and  $B = \begin{bmatrix} -2 & 1 \\ \frac{3}{2} & -\frac{1}{2} \end{bmatrix}$ . Then

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

and

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

so  $B$  is an inverse of  $A$ .

## Does every square matrix have an inverse?

No! Take e.g. the zero matrix  $\mathbf{0}_n$  (all entries of  $\mathbf{0}_n$  are equal to 0)

$$A\mathbf{0}_n = \mathbf{0}_n A = \mathbf{0}_n$$

for all  $n \times n$  matrices  $A$ : The  $(i, j)$ -entry of  $\mathbf{0}_n A$  is equal to  $\sum_{k=1}^n 0a_{kj} = 0$ .

## Does every **nonzero** square matrix have an inverse?

### Example

Does the matrix

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

have an inverse?

**No!** To see this, suppose

$$B = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

is an inverse of  $A$ . Then

$$AB = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} a & b \\ c & d \end{bmatrix} = \begin{bmatrix} c & d \\ c & d \end{bmatrix}$$

which is never equal to  $I_2$ . (Why?)

## Uniqueness of an Inverse

### Theorem

If  $A$  is a square matrix and  $B$  and  $C$  are inverses of  $A$ , then  $B = C$ .

### Proof.

Since  $B$  and  $C$  are inverses of  $A$ ,  $AB = I = BA$  and  $AC = I = CA$ . Then

$$B = BI = B(AC) = (BA)C = IC = C$$

so  $B = C$ . ◻



### Example (revisited)

For  $A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$  and  $B = \begin{bmatrix} -2 & 1 \\ \frac{3}{2} & -\frac{1}{2} \end{bmatrix}$ , we saw that

$$AB = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \text{ and } BA = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

The preceding theorem tells us that  $B$  is **the inverse** of  $A$ , rather than just **an inverse** of  $A$ .

### Definitions

Let  $A$  be a square matrix, i.e., an  $n \times n$  matrix.

- **The** inverse of  $A$ , if it exists, is denoted  $A^{-1}$ , and

$$AA^{-1} = I = A^{-1}A$$

- If  $A$  has an inverse, then we say that  $A$  is **invertible** (or **nonsingular**).

## Finding the inverse of a $2 \times 2$ matrix

### Example

Suppose that  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ . If  $ad - bc \neq 0$ , then there is a formula for  $A^{-1}$ :

$$A^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

This can easily be verified by computing the products  $AA^{-1}$  and  $A^{-1}A$ .

$$\begin{aligned} AA^{-1} &= \begin{bmatrix} a & b \\ c & d \end{bmatrix} \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \\ &= \frac{1}{ad - bc} \left( \begin{bmatrix} a & b \\ c & d \end{bmatrix} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix} \right) \\ &= \frac{1}{ad - bc} \begin{bmatrix} ad - bc & 0 \\ 0 & -bc + ad \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \end{aligned}$$

Showing that  $A^{-1}A = I_2$  is left as an exercise.

## Finding the inverse of an $n \times n$ matrix

### Problem

Suppose that  $A$  is any  $n \times n$  matrix.

- How do we know whether or not  $A^{-1}$  exists?
- If  $A^{-1}$  exists, how do we find it?

### Solution

The matrix inversion algorithm.

Although the formula for the inverse of a  $2 \times 2$  matrix is quicker and easier to use than the matrix inversion algorithm, the general formula for the inverse an  $n \times n$  matrix,  $n \geq 3$  (which we will see later), is more complicated and difficult to use than the matrix inversion algorithm. To find inverses of square matrices that are not  $2 \times 2$ , the matrix inversion algorithm is the most efficient method to use.

## The Matrix Inversion Algorithm

Let  $A$  be an  $n \times n$  matrix. To find  $A^{-1}$ , if it exists,

- take the  $n \times 2n$  matrix

$$\left[ A \mid I_n \right]$$

obtained by augmenting  $A$  with the  $n \times n$  identity matrix,  $I_n$ .

- Perform elementary row operations to transform  $\left[ A \mid I_n \right]$  into a reduced row-echelon matrix.

## Theorem (Matrix Inverses)

Let  $A$  be an  $n \times n$  matrix. Then the following conditions are equivalent.

- 1  $A$  is invertible.
- 2 the reduced row-echelon form on  $A$  is  $I$ .
- 3  $\left[ A \mid I_n \right]$  can be transformed into  $\left[ I_n \mid A^{-1} \right]$  using the Matrix Inversion Algorithm.

## Problem

Find, if possible, the inverse of  $\begin{bmatrix} 1 & 0 & -1 \\ -2 & 1 & 3 \\ -1 & 1 & 2 \end{bmatrix}$ .

## Solution

Using the matrix inversion algorithm (fill in the operations)

$$\left[ \begin{array}{ccc|ccc} 1 & 0 & -1 & 1 & 0 & 0 \\ -2 & 1 & 3 & 0 & 1 & 0 \\ -1 & 1 & 2 & 0 & 0 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 2 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 1 \end{array} \right] \rightarrow$$
$$\left[ \begin{array}{ccc|ccc} 1 & 0 & -1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & -1 & -1 & 1 \end{array} \right]$$

From this, we see that  $A$  has no inverse.

### Problem

Let  $A = \begin{bmatrix} 3 & 1 & 2 \\ 1 & -1 & 3 \\ 1 & 2 & 4 \end{bmatrix}$ . Find the inverse of  $A$ , if it exists.

### Solution (continued)

*Using the matrix inversion algorithm (fill in the operations)*

$$\begin{aligned} [A \mid I] &= \left[ \begin{array}{ccc|ccc} 3 & 1 & 2 & 1 & 0 & 0 \\ 1 & -1 & 3 & 0 & 1 & 0 \\ 1 & 2 & 4 & 0 & 0 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{ccc|ccc} 1 & -1 & 3 & 0 & 1 & 0 \\ 3 & 1 & 2 & 1 & 0 & 0 \\ 1 & 2 & 4 & 0 & 0 & 1 \end{array} \right] \rightarrow \\ &\left[ \begin{array}{ccc|ccc} 1 & -1 & 3 & 0 & 1 & 0 \\ 0 & 4 & -7 & 1 & -3 & 0 \\ 0 & 3 & 1 & 0 & -1 & 1 \end{array} \right] \rightarrow \left[ \begin{array}{ccc|ccc} 1 & -1 & 3 & 0 & 1 & 0 \\ 0 & 1 & -8 & 1 & -2 & -1 \\ 0 & 3 & 1 & 0 & -1 & 1 \end{array} \right] \rightarrow \\ &\left[ \begin{array}{ccc|ccc} 1 & 0 & -5 & 1 & -1 & -1 \\ 0 & 1 & -8 & 1 & -2 & -1 \\ 0 & 0 & 25 & -3 & 5 & 4 \end{array} \right] \rightarrow \left[ \begin{array}{ccc|ccc} 1 & 0 & -5 & 1 & -1 & -1 \\ 0 & 1 & -8 & 1 & -2 & -1 \\ 0 & 0 & 1 & -\frac{3}{25} & \frac{5}{25} & \frac{4}{25} \end{array} \right] \rightarrow \\ &\left[ \begin{array}{ccc|ccc} 1 & 0 & 0 & \frac{10}{25} & 0 & -\frac{5}{25} \\ 0 & 1 & 0 & \frac{1}{25} & -\frac{10}{25} & \frac{7}{25} \\ 0 & 0 & 1 & -\frac{3}{25} & \frac{5}{25} & \frac{4}{25} \end{array} \right] = [I \mid A^{-1}] \end{aligned}$$

### Solution (continued)

Therefore,  $A^{-1}$  exists, and

$$A^{-1} = \begin{bmatrix} \frac{10}{25} & 0 & -\frac{5}{25} \\ \frac{1}{25} & -\frac{10}{25} & \frac{7}{25} \\ -\frac{3}{25} & \frac{5}{25} & \frac{4}{25} \end{bmatrix} = \frac{1}{25} \begin{bmatrix} 10 & 0 & -5 \\ 1 & -10 & 7 \\ -3 & 5 & 4 \end{bmatrix}$$

You can check your work by computing  $AA^{-1}$  and  $A^{-1}A$ .

## Systems of Linear Equations and Inverses

Suppose that a system of  $n$  linear equations in  $n$  variables is written in matrix form as  $AX = B$ , and suppose that  $A$  is invertible.

### Example

The system of linear equations

$$\begin{aligned} 2x - 7y &= 3 \\ 5x - 18y &= 8 \end{aligned}$$

can be written in matrix form as  $AX = B$ :

$$\begin{bmatrix} 2 & -7 \\ 5 & -18 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 3 \\ 8 \end{bmatrix}$$

You can check that  $A^{-1} = \begin{bmatrix} 18 & -7 \\ 5 & -2 \end{bmatrix}$ .

### Example (continued)

Since  $A^{-1}$  exists and has the property that  $A^{-1}A = I$ , we obtain the following.

$$\begin{aligned}AX &= B \\A^{-1}(AX) &= A^{-1}B \\(A^{-1}A)X &= A^{-1}B \\IX &= A^{-1}B \\X &= A^{-1}B\end{aligned}$$

i.e.,  $AX = B$  has the **unique solution** given by  $X = A^{-1}B$ . Therefore,

$$X = A^{-1} \begin{bmatrix} 3 \\ 8 \end{bmatrix} = \begin{bmatrix} 18 & -7 \\ 5 & -2 \end{bmatrix} \begin{bmatrix} 3 \\ 8 \end{bmatrix} = \begin{bmatrix} -2 \\ -1 \end{bmatrix}$$

You should verify that  $x = -2$ ,  $y = -1$  is a solution to the system.

The last example illustrates another method for solving a system of linear equations when **the coefficient matrix is square and invertible**. Unless that coefficient matrix is  $2 \times 2$ , this is generally **NOT** an efficient method for solving a system of linear equations.

### Example

Let  $A$ ,  $B$  and  $C$  be matrices, and suppose that  $A$  is invertible.

① If  $AB = AC$ , then

$$\begin{aligned}A^{-1}(AB) &= A^{-1}(AC) \\(A^{-1}A)B &= (A^{-1}A)C \\IB &= IC \\B &= C\end{aligned}$$

② If  $BA = CA$ , then

$$\begin{aligned}(BA)A^{-1} &= (CA)A^{-1} \\B(AA^{-1}) &= C(AA^{-1}) \\BI &= CI \\B &= C\end{aligned}$$

### Problem

Find square matrices  $A$ ,  $B$  and  $C$  for which  $AB = AC$  but  $B \neq C$ .

## Inverses of Transposes and Products

### Example

Suppose  $A$  is an invertible matrix. Then

$$A^T(A^{-1})^T = (A^{-1}A)^T = I^T = I$$

and

$$(A^{-1})^T A^T = (AA^{-1})^T = I^T = I$$

This means that  $(A^T)^{-1} = (A^{-1})^T$ .

### Example

Suppose  $A$  and  $B$  are invertible  $n \times n$  matrices. Then

$$(AB)(B^{-1}A^{-1}) = A(BB^{-1})A^{-1} = AIA^{-1} = AA^{-1} = I$$

and

$$(B^{-1}A^{-1})(AB) = B^{-1}(A^{-1}A)B = B^{-1}IB = B^{-1}B = I$$

This means that  $(AB)^{-1} = B^{-1}A^{-1}$ .

# Inverses of Transposes and Products

The previous two examples prove the first two parts of the following theorem.

## Theorem

- ① If  $A$  is an invertible matrix, then  $(A^T)^{-1} = (A^{-1})^T$ .
- ② If  $A$  and  $B$  are invertible matrices, then  $AB$  is invertible and

$$(AB)^{-1} = B^{-1}A^{-1}$$

- ③ If  $A_1, A_2, \dots, A_k$  are invertible, then  $A_1A_2 \cdots A_k$  is invertible and

$$(A_1A_2 \cdots A_k)^{-1} = A_k^{-1}A_{k-1}^{-1} \cdots A_2^{-1}A_1^{-1}$$

(the third part is proved by iterating the above, or, more formally, by using mathematical induction)

## Properties of Inverses

## Theorem

- ①  $I$  is invertible, and  $I^{-1} = I$ .
- ② If  $A$  is invertible, so is  $A^{-1}$ , and  $(A^{-1})^{-1} = A$ .
- ③ If  $A$  is invertible, so is  $A^k$ , and  $(A^k)^{-1} = (A^{-1})^k$ .  
( $A^k$  means  $A$  multiplied by itself  $k$  times)
- ④ If  $A$  is invertible and  $p \in \mathbb{R}$  is nonzero, then  $pA$  is invertible, and  $(pA)^{-1} = \frac{1}{p}A^{-1}$ .



### Example

Given  $(3I - A^T)^{-1} = 2 \begin{bmatrix} 1 & 1 \\ 2 & 3 \end{bmatrix}$ , we wish to find the matrix  $A$ . Taking inverses of both sides of the equation:

### Example (continued)

## Problem

True or false? Justify your answer.

If  $A^3 = 4I$ , then  $A$  is invertible.

## Solution

## A Fundamental Result

### Theorem

Let  $A$  be an  $n \times n$  matrix, and let  $X, B$  be  $n \times 1$  vectors. The following conditions are equivalent.

- ① The rank of  $A$  is  $n$ .
- ②  $A$  can be transformed to  $I_n$  by elementary row operations.
- ③  $A$  is invertible.
- ④ There exists an  $n \times n$  matrix  $C$  with the property that  $CA = I_n$ .
- ⑤ The system  $AX = B$  has a unique solution  $X$  for any choice of  $B$ .
- ⑥  $AX = 0$  has only the trivial solution,  $X = 0$ .
- ⑦ There exists an  $n \times n$  matrix  $C$  with the property that  $AC = I_n$ .

### Proof of Theorem:

(1)  $\Rightarrow$  (2) The rank of  $A$  is the number of leading 1s in the RREF of  $A$ . Since the size of  $A$  is  $n \times n$ ,  $\text{rank}(A) = n$  is equivalent to  $A$  being row-equivalent to  $I_n$ .

(2)  $\Rightarrow$  (3): Matrix inversion algorithm.

(3)  $\Rightarrow$  (4):  $C = A^{-1}$ .

(4)  $\Rightarrow$  (5):  $X = CB$ .

(5)  $\Rightarrow$  (6): Take  $B = 0$ .

(6)  $\Rightarrow$  (1): If  $\text{rank}$  of  $A$  is  $< n$ , then there are non-leading variables in the RREF of  $[A|0]$ . Hence  $AX = 0$  has infinitely many solutions.

(4)  $\Leftrightarrow$  (7):  $CA = I$  if and only if  $A^T C^T = I$ ; hence (4) for  $A$  is equivalent to (7) for  $A^T$ .

We already know that  $A^{-1}$  exists if and only if  $(A^T)^{-1}$  exists.

The following is an important and useful consequence of the theorem.

### Theorem

*If  $A$  and  $B$  are  $n \times n$  matrices such that  $AB = I$ , then  $BA = I$ . Furthermore,  $A$  and  $B$  are invertible, with  $B = A^{-1}$  and  $A = B^{-1}$ .*

### Important Fact

In the second Theorem, it is essential that the matrices be square.

## Theorem

If  $A$  and  $B$  are matrices such that  $AB = I$  and  $BA = I$ , then  $A$  and  $B$  are square matrices (of the same size).

## Example

Let  $A = \begin{bmatrix} 1 & 1 & 0 \\ -1 & 4 & 1 \end{bmatrix}$  and  $B = \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 1 & 1 \end{bmatrix}$ . Then

$$AB = \begin{bmatrix} 1 & 1 & 0 \\ -1 & 4 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_2$$

and

$$BA = \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 0 \\ -1 & 4 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \\ 0 & 5 & 1 \end{bmatrix} \neq I_3$$

This example illustrates why “an inverse” of a non-square matrix doesn’t make sense. If  $A$  is  $m \times n$  and  $B$  is  $n \times m$ , where  $m \neq n$ , then even if  $AB = I$ , it will never be the case that  $BA = I$ .