# 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

These lecture slides were originally developed by Karen Seyffarth of the University of Calgary. Edits, additions, and revisions have been made to these notes by the editorial team at Lyryx Learning to accompany their text A First Course in Linear Algebra based on K. Kuttler's original text.

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#### **Definitions**

Let m and n be positive integers.

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

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



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- **3** 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

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



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### Example

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

$$A + B =$$



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

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- There exists an  $m \times n$  matrix -A such that A + (-A) = 0. (existence of an additive inverse).

# Scalar Multiplication

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.

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$$3A =$$

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

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- k(A + B) = kA + kB. (scalar multiplication distributes over matrix addition).
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- k(A + B) = kA + kB. (scalar multiplication distributes over matrix addition).
- (addition distributes over scalar multiplication).
- $\bullet$  1A = A. (existence of a multiplicative identity).



### Example

$$2\left[\begin{array}{cc} -1 & 0 \\ 1 & 1 \end{array}\right] + 4\left[\begin{array}{cc} -2 & 1 \\ 3 & 0 \end{array}\right] - \left[\begin{array}{cc} 6 & 8 \\ 1 & -1 \end{array}\right] =$$

### Example

$$2\left[\begin{array}{cc}-1 & 0 \\ 1 & 1\end{array}\right]+4\left[\begin{array}{cc}-2 & 1 \\ 3 & 0\end{array}\right]-\left[\begin{array}{cc}6 & 8 \\ 1 & -1\end{array}\right]=$$

#### **Problem**

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

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







### Example

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#### Solution

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

$$= 2(9A - 9B + 14B - 7A) - 2(6B + 3A - 2A - 6B - 5A - 5B)$$

$$= 2(2A+5B)-2(-4A-5B)$$

$$= 12A + 20B$$

#### **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

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

```
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 \vdots \vdots a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m
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 $\vdots$   $\vdots$   $\vdots$   $\vdots$   $\vdots$   $\vdots$   $a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m$ 

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.

$$2x_1 + 4x_2 - 3x_3 = -6$$
  
 $- x_2 + 5x_3 = 0$   
 $x_1 + x_2 + 4x_3 = 1$ 

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### Solution

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

#### Definition

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

$$X = \left[ \begin{array}{c} x_1 \\ x_2 \\ \vdots \\ x_n \end{array} \right]$$

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

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.

#### **Problem**

Compute the product AX for

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

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### Solution

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

#### **Problem**

Compute AY for

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

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Compute AY for

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### Solution

$$AY =$$

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 $\vdots$   $\vdots$   $\vdots$   $\vdots$   $\vdots$   $\vdots$   $a_{m1}x_1 + a_{m2}x_2 + \cdots + a_{mn}x_n = b_m$ 

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

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

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

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Express the following system of linear equations in matrix form.

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### Solution

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## 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.

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- 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.
- 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  $x_1, x_2, \ldots, x_n$  are a solution to the vector equation

$$x_1A_1+x_2A_2+\cdots x_nA_n=B$$

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



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#### Proof.

(a) One first checks that  $(x_1, \ldots, x_n)$  is a solution to the original system if

and only if 
$$X = \begin{bmatrix} \frac{x_1}{x_2} \\ \vdots \\ \frac{x_n}{x_n} \end{bmatrix}$$
 is a solution to  $AX = B$ .

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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:

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

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



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.





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

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Since there are infinitely many solutions ( $x_4$  is assigned a parameter), choose any value for  $x_4$ .

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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  $B = \begin{bmatrix} B_1 & B_2 & \cdots & B_p \end{bmatrix}$  be an  $n \times p$  matrix, whose columns are  $B_1, B_2, \ldots, B_p$ . The product of A and B is the matrix

$$AB = A \begin{bmatrix} B_1 & B_2 & \cdots & B_p \end{bmatrix} = \begin{bmatrix} AB_1 & AB_2 & \cdots & AB_p \end{bmatrix}$$

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.)

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

(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

$$\left[\begin{array}{ccc} -1 & 0 & 3 \\ 2 & -1 & 1 \end{array}\right] \left[\begin{array}{cccc} -1 & 1 & 2 \\ 0 & -2 & 4 \\ 1 & 0 & 0 \end{array}\right]$$

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

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

Find the product AB of matrices

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

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### 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

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

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Note that the product

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

does not exist.

# Multiplication by the Zero Matrix

## Example

Compute the product A0 for the matrix

$$A = \left[ \begin{array}{cc} 1 & 2 \\ 3 & 4 \end{array} \right]$$

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

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

#### Solution

In this product, we compute

$$\left[\begin{array}{cc} 1 & 2 \\ 3 & 4 \end{array}\right] \left[\begin{array}{cc} 0 & 0 \\ 0 & 0 \end{array}\right] = \left[\begin{array}{cc} 0 & 0 \\ 0 & 0 \end{array}\right]$$

Hence. A0 = 0.

Given matrices A and B, is AB = BA?





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

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

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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.





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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?





#### **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.



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

$$AB = \left[ \begin{array}{rrrr} 7 & -5 & 4 & -6 \\ -3 & 3 & -6 & 0 \\ -11 & 7 & -2 & 12 \end{array} \right]$$





#### 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





Let

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

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







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 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!





Let

$$P = \left[ egin{array}{cc} 1 & 0 \\ 2 & -1 \end{array} 
ight] \ ext{and} \ Q = \left[ egin{array}{cc} -1 & 1 \\ 0 & 3 \end{array} 
ight]$$

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





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

#### Solution

$$PQ = \left[ \begin{array}{cc} -1 & 1 \\ -2 & -1 \end{array} \right]$$

$$QP = \left| \begin{array}{cc} 1 & -1 \\ 6 & -3 \end{array} \right|$$

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  $AB \neq BA$ .





Let

$$U = \left[ \begin{array}{cc} 2 & 0 \\ 0 & 2 \end{array} \right]$$
 and  $V = \left[ \begin{array}{cc} 1 & 2 \\ 3 & 4 \end{array} \right]$ 

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

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- Does UV exist? If so, compute it.
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$$UV = \left[ \begin{array}{cc} 2 & 4 \\ 6 & 8 \end{array} \right]$$



Let

$$U = \left[ \begin{array}{cc} 2 & 0 \\ 0 & 2 \end{array} \right] \text{ and } V = \left[ \begin{array}{cc} 1 & 2 \\ 3 & 4 \end{array} \right]$$

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

### Solution

$$UV = \left[ \begin{array}{cc} 2 & 4 \\ 6 & 8 \end{array} \right]$$

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

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





#### **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.



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• A(B+C) = AB + AC. (matrix multiplication distributes over matrix addition).





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- **3** A(BC) = (AB) C. (matrix multiplication is associative).



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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.

- (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).
- (AB) = (rA)B = A(rB).





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



### **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.

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# 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.



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.



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## Proof.





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### Proof.

$$(AB)C = A(BC)$$
 (matrix multiplication is associative)





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### Proof.

$$(AB)C = A(BC)$$
 (matrix multiplication is associative)  
=  $A(CB)$  (B commutes with C)



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.

$$(AB)C = A(BC)$$
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=  $A(CB)$  (B commutes with C)  
=  $(AC)B$  (matrix multiplication is associative)



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.

$$(AB)C = A(BC)$$
 (matrix multiplication is associative)  
=  $A(CB)$  ( $B$  commutes with  $C$ )  
=  $(AC)B$  (matrix multiplication is associative)



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.

$$(AB)C = A(BC)$$
 (matrix multiplication is associative)

- = A(CB) (B commutes with C)
- = (AC)B (matrix multiplication is associative)
- = (CA)B (A commutes with C)
- = C(AB) (matrix multiplication is associative)



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.

$$(AB)C = A(BC)$$
 (matrix multiplication is associative)

= A(CB) (B commutes with C)

= (AC)B (matrix multiplication is associative)

= (CA)B (A commutes with C)

= C(AB) (matrix multiplication is associative)

Therefore, AB commutes with C.





If A is an  $m \times n$  matrix, then its transpose, denoted  $A^T$ , is the  $n \times m$  whose  $i^{th}$  row is the  $i^{th}$  column of A,  $1 \le i \le n$ ; 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.

#### LECTURE 2



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$$(A + B)^T = A^T + B^T$$

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

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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.

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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!



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



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# Solution



# Symmetric Matrices

### **Definition**

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

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### **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.

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### 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**

$$\left[\begin{array}{ccc} 2 & -3 \\ -3 & 17 \end{array}\right], \left[\begin{array}{cccc} -1 & 0 & 5 \\ 0 & 2 & 11 \\ 5 & 11 & -3 \end{array}\right], \left[\begin{array}{ccccc} 0 & 2 & 5 & -1 \\ 2 & 1 & -3 & 0 \\ 5 & -3 & 2 & -7 \\ -1 & 0 & -7 & 4 \end{array}\right]$$

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



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#### Definition

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

# Example (Skew Symmetric Matrices)

$$\left[\begin{array}{ccc} 0 & 2 \\ -2 & 0 \end{array}\right], \left[\begin{array}{cccc} 0 & 9 & 4 \\ -9 & 0 & -3 \\ -4 & 3 & 0 \end{array}\right]$$

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## **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  $n \ge 2$ , the  $n \times n$  identity matrix, denoted  $l_n$ , is the matrix having ones on its main diagonal and zeros elsewhere, and is defined for all  $n \ge 2$ .

## Example

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

# The $n \times n$ Identity Matrix

### Definition

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## 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  $E_i$  the j<sup>th</sup> column of  $I_n$ .

## Example

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  $I_mA = A$ .

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#### Proof

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

$$\left[\begin{array}{cccc} a_{i1} & a_{i2} & \cdots & a_{ij} & \cdots & a_{in} \end{array}\right] E_j = a_{ij}$$

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  $I_mA$  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.

Instead of  $AI_n$  and  $I_mA$  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

$$AI = A$$
 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  $BA = I_n$ .

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### 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

$$AB = \left[ \begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array} \right]$$

and

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

so B is an inverse of A.



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

$$\textit{A}0_{n}=0_{n}\textit{A}=O_{n}$$

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





# Example

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is an inverse of A. Then

$$AB = \left[ \begin{array}{cc} 0 & 1 \\ 0 & 1 \end{array} \right] \left[ \begin{array}{cc} a & b \\ c & d \end{array} \right] = \left[ \begin{array}{cc} c & d \\ c & d \end{array} \right]$$

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

#### Theorem

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

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#### 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

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

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

### **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).



## Example

Suppose that 
$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$
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## Example

Suppose that 
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$$AA^{-1} = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$
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$$= \frac{1}{ad - bc} \begin{bmatrix} ad - bc & 0 \\ 0 & -bc + ad \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

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



### Problem

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

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

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### Solution

The matrix inversion algorithm.



#### **Problem**

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### 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

$$[A \mid I_n]$$

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

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

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## Theorem (Matrix Inverses)

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

- A is invertible.
- 2 the reduced row-echelon form on A is I.
- **3**  $\begin{bmatrix} A \mid I_n \end{bmatrix}$  can be transformed into  $\begin{bmatrix} I_n \mid A^{-1} \end{bmatrix}$  using the Matrix Inversion Algorithm.

### Problem

Find, if possible, the inverse of  $\left[ \begin{array}{ccc} 1 & 0 & -1 \\ -2 & 1 & 3 \\ -1 & 1 & 2 \end{array} \right].$ 





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Using the matrix inversion algorithm (fill in the operations)





#### **Problem**

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Using the matrix inversion algorithm (fill in the operations)

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



#### Problem

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

#### Solution

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

#### **Problem**

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Using the matrix inversion algorithm (fill in the operations)

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

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.





$$\left[\begin{array}{c|cccc} A & I \end{array}\right] = \left[\begin{array}{cccccc} 3 & 1 & 2 & 1 & 0 & 0 \\ 1 & -1 & 3 & 0 & 1 & 0 \\ 1 & 2 & 4 & 0 & 0 & 1 \end{array}\right]$$

$$\left[\begin{array}{c|ccccc} A & I \end{array}\right] = \left[\begin{array}{cccccc} 3 & 1 & 2 & 1 & 0 & 0 \\ 1 & -1 & 3 & 0 & 1 & 0 \\ 1 & 2 & 4 & 0 & 0 & 1 \end{array}\right] \rightarrow \left[\begin{array}{cccccccc} 1 & -1 & 3 & 0 & 1 & 0 \\ 3 & 1 & 2 & 1 & 0 & 0 \\ 1 & 2 & 4 & 0 & 0 & 1 \end{array}\right] \rightarrow$$

$$\begin{bmatrix} 1 & -1 & 3 & 0 & 1 & 0 \\ 0 & 4 & -7 & 1 & -3 & 0 \\ 0 & 3 & 1 & 0 & -1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1 & 3 & 0 & 1 & 0 \\ 0 & 1 & -8 & 1 & -2 & -1 \\ 0 & 3 & 1 & 0 & -1 & 1 \end{bmatrix} \rightarrow$$

$$\begin{bmatrix} 1 & 0 & -5 & 1 & -1 & -1 \\ 0 & 1 & -8 & 1 & -2 & -1 \\ 0 & 0 & 25 & -3 & 5 & 4 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -5 & 1 & -1 & -1 \\ 0 & 1 & -8 & 1 & -2 & -1 \\ 0 & 0 & 1 & -\frac{3}{25} & \frac{5}{25} & \frac{4}{25} \end{bmatrix} \rightarrow$$

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

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

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

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

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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.



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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

$$2x - 7y = 3$$
$$5x - 18y = 8$$

can be written in matrix form as AX = B:

$$\left[\begin{array}{cc} 2 & -7 \\ 5 & -18 \end{array}\right] \left[\begin{array}{c} x \\ y \end{array}\right] = \left[\begin{array}{c} 3 \\ 8 \end{array}\right]$$

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



$$AX = B$$

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



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$$IX = A^{-1}B$$

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

$$AX = B$$

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

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

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

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

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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**.

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.

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

• If AB = AC, then

• If 
$$AB = AC$$
, then

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





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





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② If BA = CA, then



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$$(BA)A^{-1} = (CA)A^{-1}$$

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$$B(AA^{-1}) = C(AA^{-1})$$

$$BI = CI$$

$$B = C$$

#### **Problem**

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



### Example

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#### Theorem

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- **1** If A is an invertible matrix, then  $(A^T)^{-1} = (A^{-1})^T$ .
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3 If  $A_1, A_2, \ldots, A_k$  are invertible, then  $A_1 A_2 \cdots A_k$  is invertible and

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

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





#### Theorem

**1** I is invertible, and  $I^{-1} = I$ .



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- **1** If A is invertible and  $p \in \mathbb{R}$  is nonzero, then pA is invertible, and  $(pA)^{-1} = \frac{1}{p}A^{-1}$ .





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





































True or false? Justify your answer.

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### 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.

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- The rank of A is n.
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- **1** The system AX = B has a unique solution X for any choice of B.
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- There exists an  $n \times n$  matrix C with the property that  $AC = I_n$ .

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

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



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#### 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  $A = B^{-1}$ .

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### 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).





Let 
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 and  $B = \begin{bmatrix} 1 & 0 \\ 0 & 0 \\ 1 & 1 \end{bmatrix}$ .

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$$BA = \left[ \begin{array}{cc} 1 & 0 \\ 0 & 0 \\ 1 & 1 \end{array} \right] \left[ \begin{array}{ccc} 1 & 1 & 0 \\ -1 & 4 & 1 \end{array} \right]$$





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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.

