

# Math 221: LINEAR ALGEBRA

## Chapter 2. Matrix Algebra

### §2-2. Equations, Matrices, and Transformations

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Emory University, 2021 Spring

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<sup>1</sup>Slides are adapted from those by Karen Seyffarth from University of Calgary.



# Linear Algebra with Applications

## Lecture Notes

### Current Lecture Notes Revision: Version 2018 — Revision B

These lecture notes 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 [Linear Algebra with Applications](#) based on W. K. Nicholson's original text.

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- Ilijas Farah, York University

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Vectors

Matrix Vector Multiplication

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Rotations in  $\mathbb{R}^2$

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## Vectors

Matrix Vector Multiplication

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# 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  $\vec{x}$  is a **row vector** of size  $1 \times n$ , and  $\vec{y}$  is a **column vector** of size  $m \times 1$ , then we write

$$\vec{x} = \begin{bmatrix} x_1 & x_2 & \cdots & x_n \end{bmatrix} \quad \text{and} \quad \vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{bmatrix}$$

## Definition ( Vector form of a system of linear equations )

Consider the system of linear equations

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



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

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

## Problem

Express the following system of linear equations in vector form:

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

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## Solution

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

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Vectors

**Matrix Vector Multiplication**

The Dot Product

Transformations

Rotations in  $\mathbb{R}^2$



# Matrix vector multiplication

## Definition

Let  $A = [a_{ij}]$  be an  $m \times n$  matrix with columns  $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n$ , written  $A = \begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \cdots & \vec{a}_n \end{bmatrix}$ , and let  $\vec{x}$  be an  $n \times 1$  column vector,

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

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

Then the product of matrix  $A$  and (column) vector  $\vec{x}$  is the  $m \times 1$  column vector given by

$$\begin{bmatrix} \vec{a}_1 & \vec{a}_2 & \cdots & \vec{a}_n \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1 \vec{a}_1 + x_2 \vec{a}_2 + \cdots + x_n \vec{a}_n = \sum_{j=1}^n x_j \vec{a}_j$$

that is,  $A\vec{x}$  is a linear combination of the columns of  $A$ .

## Problem

Compute the product  $A\vec{x}$  for

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



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### Solution

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

## Problem

Compute  $A\vec{y}$  for

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

### Problem

Compute  $A\vec{y}$  for

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### Solution

$$A\vec{y} = 2 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + (-1) \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 2 \\ 0 \\ 3 \end{bmatrix} + 4 \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 9 \\ 12 \end{bmatrix}$$

## Definition ( Matrix form of a system of linear equations )

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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 expressed as a **matrix equation**

$$A\vec{x} = \vec{b},$$

where  $A$  is the coefficient matrix,  $\vec{b}$  is the constant matrix, and  $\vec{x}$  is the matrix of variables.

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## Solution

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

## Theorem

1. Every system of  $m$  linear equations in  $n$  variables can be written in the form  $A\vec{x} = \vec{b}$  where  $A$  is the coefficient matrix,  $\vec{x}$  is the matrix of variables, and  $\vec{b}$  is the constant matrix.



### Theorem (continued)

2. The system  $A\vec{x} = \vec{b}$  is consistent (i.e., has at least one solution) if and only if  $\vec{b}$  is a linear combination of the columns of  $A$ .

### Theorem (continued)

3. The vector  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$  is a solution to the system  $A\vec{x} = \vec{b}$  if and only if  $x_1, x_2, \dots, x_n$  are a solution to the vector equation

$$x_1\vec{a}_1 + x_2\vec{a}_2 + \cdots x_n\vec{a}_n = \vec{b}$$

where  $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n$  are the columns of  $A$ .

## Problem

Let

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

Express  $\vec{b}$  as a linear combination of the columns  $\vec{a}_1, \vec{a}_2, \vec{a}_3, \vec{a}_4$  of  $A$ , or show that this is impossible.

## Solution

Solve the system  $A\vec{x} = \vec{b}$  where  $\vec{x}$  is a column vector with four entries.

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Solve the system  $A\vec{x} = \vec{b}$  where  $\vec{x}$  is a column vector with four entries. Do so by putting the augmented matrix  $\left[ A \mid \vec{b} \right]$  in reduced row-echelon form.

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

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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

$$\vec{b} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{7} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} - \frac{5}{7} \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} + \frac{3}{7} \begin{bmatrix} 2 \\ 0 \\ 3 \end{bmatrix} = \frac{1}{7}\vec{a}_1 - \frac{5}{7}\vec{a}_2 + \frac{3}{7}\vec{a}_3 + 0\vec{a}_4.$$





## Remark

The problem may ask to find **all possible** linear combinations of the columns  $\vec{a}_1, \vec{a}_2, \vec{a}_3, \vec{a}_4$  of  $A$ .

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This is equivalent to find all solutions to the corresponding system of linear equations:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} \frac{1}{7} - s \\ -\frac{5}{7} - s \\ \frac{3}{7} + s \\ s \end{bmatrix}$$

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Hence, all possible linear combinations are:

$$\vec{b} = \left(\frac{1}{7} - s\right) \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} - \left(\frac{5}{7} + s\right) \begin{bmatrix} 0 \\ -1 \\ 1 \end{bmatrix} + \left(\frac{3}{7} + s\right) \begin{bmatrix} 2 \\ 0 \\ 3 \end{bmatrix} + s \begin{bmatrix} -1 \\ 1 \\ 1 \end{bmatrix}$$

## Theorem

Let  $A$  and  $B$  be  $m \times n$  matrices, and let  $\vec{x}$  and  $\vec{y}$  be  $n$ -vectors in  $\mathbb{R}^n$ . Then:

1.  $A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y}$ .
2.  $A(a\vec{x}) = a(A\vec{x}) = (aA)\vec{x}$  for all scalars  $a$ .
3.  $(A + B)\vec{x} = A\vec{x} + B\vec{x}$ .

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This provides a useful way to describe the solutions to a system  $A\vec{x} = \vec{b}$ .

Structure of solutions:

General solution = Sol. to the Homog. Eq. + A Particular Solution.

$$A\vec{x} = A(\vec{x}_0 + \vec{x}_1) = \underbrace{A\vec{x}_0}_{\vec{x}_0: \text{homogeneous sol.}} + \underbrace{A\vec{x}_1}_{\vec{x}_1: \text{particular sol.}} = \vec{0} + \vec{b} = \vec{b}.$$

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# The Dot Product

## Definition

If  $(a_1, a_2, \dots, a_n)$  and  $(b_1, b_2, \dots, b_n)$  are two ordered  $n$ -tuples, their **dot product** is defined to be the number

$$a_1b_1 + a_2b_2 + \cdots + a_nb_n$$

obtained by multiplying corresponding entries and adding the results.



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obtained by multiplying corresponding entries and adding the results.

This give an alternative way to carry out the matrix-vector product  $A\vec{x}$ .

$$\begin{array}{ccc} A & \vec{x} & A\vec{x} \\ \left[ \begin{array}{c} \text{row } i \end{array} \right] & \left[ \begin{array}{c} \text{entry } i \end{array} \right] & = \left[ \begin{array}{c} \text{entry } i \end{array} \right] \end{array}$$

The diagram shows the matrix-vector product  $A\vec{x}$  as a dot product. On the left, a matrix  $A$  is represented by a row vector (a horizontal blue oval with a right-pointing arrow) labeled "row i". In the middle, a vector  $\vec{x}$  is represented by a column vector (a vertical blue oval with a downward-pointing arrow) labeled "entry i". An equals sign follows, and on the right, the result  $A\vec{x}$  is represented by a single entry (a small blue circle) labeled "entry i".

$$A\vec{x}$$

$$||$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$$

$$||$$

$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ a_{31} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ a_{32} \end{bmatrix} + x_3 \begin{bmatrix} a_{13} \\ a_{23} \\ a_{33} \end{bmatrix} + x_4 \begin{bmatrix} a_{14} \\ a_{24} \\ a_{34} \end{bmatrix} \qquad \text{(Def.)}$$

$$||$$

$$\begin{bmatrix} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + a_{14}x_4 \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + a_{24}x_4 \\ a_{31}x_1 + a_{32}x_2 + a_{33}x_3 + a_{34}x_4 \end{bmatrix} \qquad \text{(Alternative)}$$

## Problem

$$\text{If } A = \begin{bmatrix} 1 & 0 & 2 & -1 \\ 2 & -1 & 0 & 1 \\ 3 & 1 & 3 & 1 \end{bmatrix} \text{ and } \vec{x} = \begin{bmatrix} 2 \\ -1 \\ 1 \\ 4 \end{bmatrix}, \text{ compute } A\vec{x}.$$


## Problem

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## Solution

The entries of  $A\vec{x}$  are the dot products of the rows of  $A$  with  $\vec{x}$ :

$$\begin{aligned} A\vec{x} &= \begin{bmatrix} 1 & 0 & 2 & -1 \\ 2 & -1 & 0 & 1 \\ 3 & 1 & 3 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ -1 \\ 1 \\ 4 \end{bmatrix} \\ &= \begin{bmatrix} 1 \cdot 2 & + & 0(-1) & + & 2 \cdot 1 & + & (-1)4 \\ 2 \cdot 2 & + & (-1)(-1) & + & 0 \cdot 1 & + & 1 \cdot 4 \\ 3 \cdot 2 & + & 1(-1) & + & 3 \cdot 1 & + & 1 \cdot 4 \end{bmatrix} = \begin{bmatrix} 0 \\ 9 \\ 12 \end{bmatrix}. \end{aligned}$$

Of course, this agrees with the outcome of the previous example. 

### Definition ( Identity Matrix )

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## Example

The first few identity matrices are

$$I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad I_4 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}, \quad \dots$$

## Problem

Show that  $I_n \vec{x} = \vec{x}$  for each  $n$ -vector  $\vec{x}$  in  $\mathbb{R}^n$ ,  $n \geq 1$ .


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## Solution

We verify the case  $n = 4$ . Given the 4-vector  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}$  the dot product rule gives

$$I_4 \vec{x} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} x_1 + 0 + 0 + 0 \\ 0 + x_2 + 0 + 0 \\ 0 + 0 + x_3 + 0 \\ 0 + 0 + 0 + x_4 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \vec{x}.$$

In general,  $I_n \vec{x} = \vec{x}$  because entry  $k$  of  $I_n \vec{x}$  is the dot product of row  $k$  of  $I_n$  with  $\vec{x}$ , and row  $k$  of  $I_n$  has 1 in position  $k$  and zeros elsewhere. 



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# Transformations

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- ▶ We have already used  $\mathbb{R}$  to denote the set of **real numbers**.
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- ▶ In general, we write  $\mathbb{R}^n$  for the set of all **column vectors of length  $n$** .



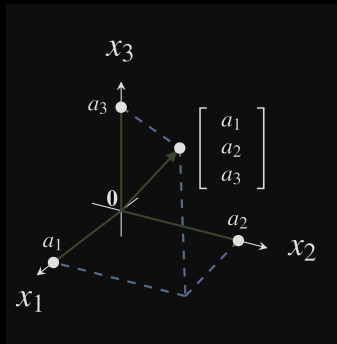
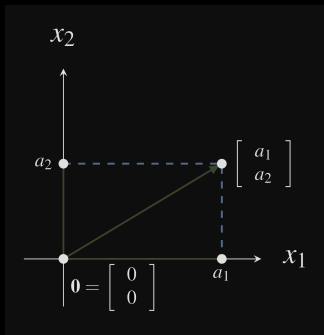
# Transformations

## Notation and Terminology

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## $\mathbb{R}^2$ and $\mathbb{R}^3$

Vectors in  $\mathbb{R}^2$  and  $\mathbb{R}^3$  have convenient geometric interpretations as position vectors of points in the 2-dimensional (Cartesian) plane and in 3-dimensional space, respectively.



## Definition (Transformations)

A **transformation** is a function  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , sometimes written  $\mathbb{R}^n \xrightarrow{T} \mathbb{R}^m$ , and is called a **transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$** .

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What do we mean by a function?

Informally, a function  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a rule that, for each vector in  $\mathbb{R}^n$ , assigns exactly one vector of  $\mathbb{R}^m$

We use the notation  $T(\vec{x})$  to mean the transformation  $T$  applied to the vector  $\vec{x}$ .

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## Definition

If  $T$  acts by matrix multiplication of a matrix  $A$  (such as the previous example), we call  $T$  a **matrix transformation**, and write  $T_A(\vec{x}) = A\vec{x}$ .

### Definition ( Equality of Transformations )

Suppose  $S : \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  are transformations. Then  $S = T$  if and only if  $S(\vec{x}) = T(\vec{x})$  for every  $\vec{x} \in \mathbb{R}^n$ .



Example ( Specifying the action of a transformation )

$T : \mathbb{R}^3 \rightarrow \mathbb{R}^4$  defined by

$$T \begin{bmatrix} a \\ b \\ c \end{bmatrix} = \begin{bmatrix} a + b \\ b + c \\ a - c \\ c - b \end{bmatrix}$$

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is a transformation that **transforms** the vector  $\begin{bmatrix} 1 \\ 4 \\ 7 \end{bmatrix}$  in  $\mathbb{R}^3$  into the vector

$$T \begin{bmatrix} 1 \\ 4 \\ 7 \end{bmatrix} = \begin{bmatrix} 1 + 4 \\ 4 + 7 \\ 1 - 7 \\ 7 - 4 \end{bmatrix} = \begin{bmatrix} 5 \\ 11 \\ -6 \\ 3 \end{bmatrix}.$$

### Example ( Transformation by matrix multiplication )

Consider the matrix  $A = \begin{bmatrix} 1 & 2 & 0 \\ 2 & 1 & 0 \end{bmatrix}$ . By matrix multiplication, A

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Copyright

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Rotations in  $\mathbb{R}^2$





## Rotations in $\mathbb{R}^2$

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Let  $A$  be an  $m \times n$  matrix. The transformation  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  defined by

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### Definition

The transformation

$$R_\theta : \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

denotes **counterclockwise rotation** about the origin through an angle of  $\theta$ .

### Example (Rotation through $\pi$ )

We denote by

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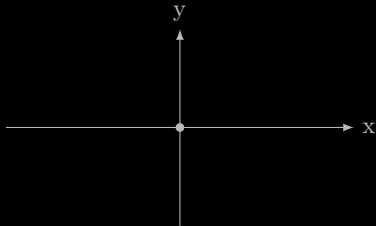
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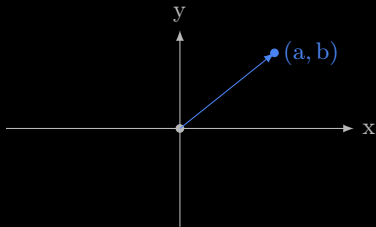


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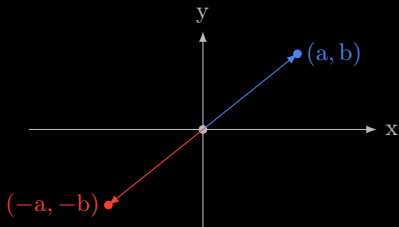


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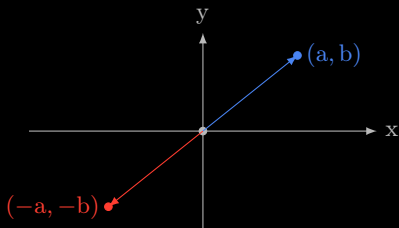


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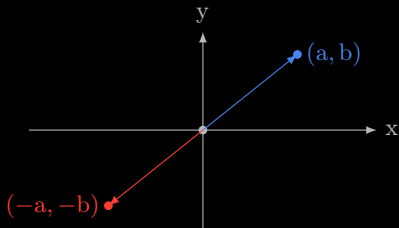
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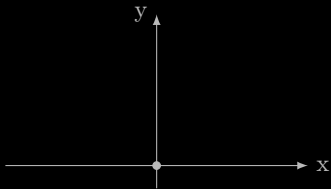
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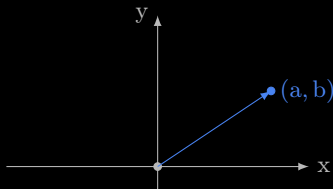


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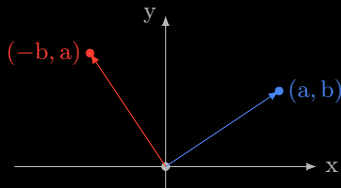


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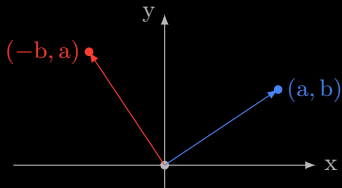


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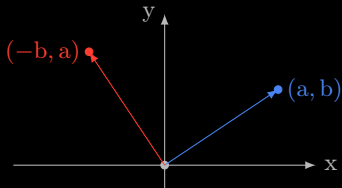
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### Remark

In general, the rotation (counterclockwise) about the origin for an angle  $\theta$  is

$$\mathbf{R}_\theta = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$

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