

# Math 221: LINEAR ALGEBRA

## Chapter 2. Matrix Algebra

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

Le Chen<sup>1</sup>

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

Vectors

Matrix Vector Multiplication

The Dot Product

Transformations

Rotations in  $\mathbb{R}^2$

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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

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

## 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:

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



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## Solution

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

Vectors

**Matrix Vector Multiplication**

The Dot Product

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# 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  $A = [ \vec{a}_1 \quad \vec{a}_2 \quad \cdots \quad \vec{a}_n ]$ , 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

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

## 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

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

## Solution

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

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

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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.

## Problem

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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

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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  $x_4$ .

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

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  $i$  of a dark blue rounded rectangle with a white arrow pointing to the right. This row is enclosed in square brackets. To its right is a column vector  $\vec{x}$ , represented by a dark blue rounded rectangle with a white arrow pointing downwards, also enclosed in square brackets. An equals sign follows, leading to the result  $A\vec{x}$ , which is a single dark blue circle with a white arrow pointing downwards, enclosed in square brackets. The label 'row i' is positioned below the first row, and 'entry i' is positioned below the resulting circle.

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

$A\vec{x}$  $\parallel$ 

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

 $\parallel$ 

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

 $\parallel$ 

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



## Problem

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

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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. ■

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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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- ▶ In general, we write  $\mathbb{R}^n$  for the set of all **column vectors of length  $n$** .

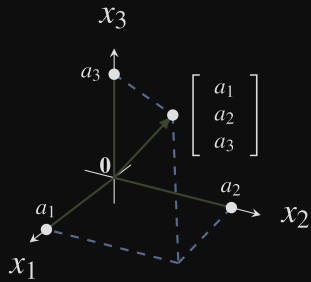
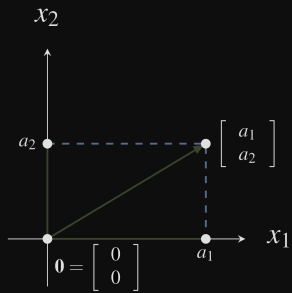
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## $\mathbb{R}^2$ and $\mathbb{R}^3$

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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## 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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Vectors

Matrix Vector Multiplication

The Dot Product

Transformations

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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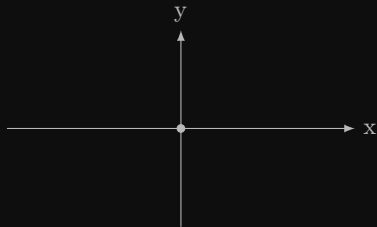
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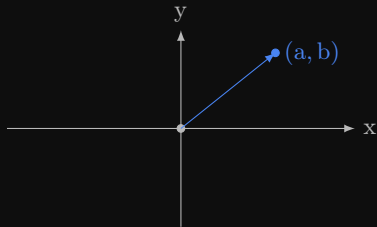


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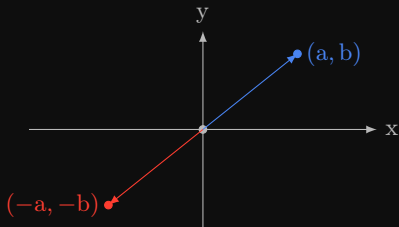


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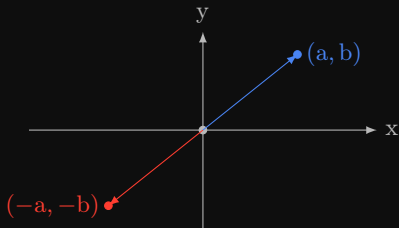


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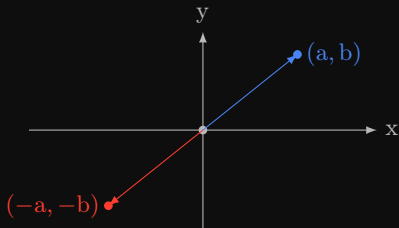
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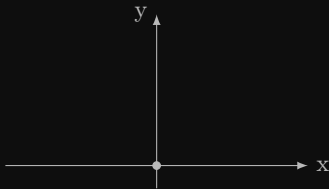
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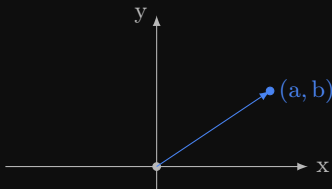


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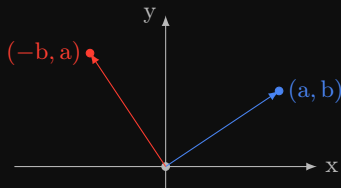


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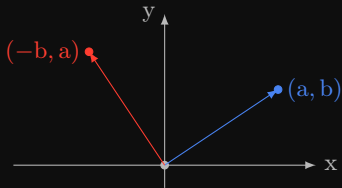


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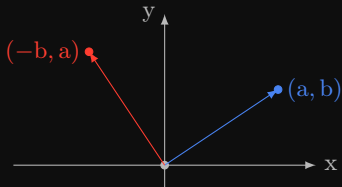
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## Remark

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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