

# Vectors as Ordered Lists or $n$ -tuples

We will **temporarily** use the word “vector” to refer to an ordered list of numbers.

## Definition

The set of all  $n$ -tuples of real numbers is called  $\mathbb{R}^n$ .

Elements of  $\mathbb{R}^n$  are usually **represented** as  $n \times 1$  column vectors ( $n \times 1$  matrices).

The vector whose entries are all zero is called the **zero vector** and is denoted by **0**.

Equality of vectors in  $\mathbb{R}^n$  and the operations of scalar multiplication and vector addition in  $\mathbb{R}^n$  are defined entry by entry just as in  $\mathbb{R}^2$ .

# Algebraic Properties of $\mathbb{R}^n$

For all  $\mathbf{u}, \mathbf{v}, \mathbf{w} \in \mathbb{R}^n$  and all scalars  $c$  and  $d$ ,

- $\mathbf{u} + \mathbf{v} = \mathbf{v} + \mathbf{u}$

- $\mathbf{u} + (\mathbf{v} + \mathbf{w}) = (\mathbf{u} + \mathbf{v}) + \mathbf{w}$

- $\mathbf{u} + \mathbf{0} = \mathbf{0} + \mathbf{u} = \mathbf{u}.$

- $\mathbf{u} + (-\mathbf{u}) = -\mathbf{u} + \mathbf{u} = \mathbf{0}$  (where  $-\mathbf{u}$  denotes  $(-1)\mathbf{u}$ )

- $c(\mathbf{u} + \mathbf{v}) = c\mathbf{u} + c\mathbf{v}$

- $(c + d)\mathbf{u} = c\mathbf{u} + d\mathbf{u}$

- $c(d\mathbf{u}) = (cd)\mathbf{u}$

- $1\mathbf{u} = \mathbf{u}$

## Definition

Given vectors  $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p \in \mathbb{R}^n$  and given scalars  $c_1, c_2, \dots, c_p$ , the vector  $\mathbf{y}$  defined by

$$\mathbf{y} = c_1 \mathbf{v}_1 + \dots + c_p \mathbf{v}_p$$

is called a *linear combination* of  $\mathbf{v}_1, \dots, \mathbf{v}_p$  with *weights* or *coefficients*  $c_1, \dots, c_p$ .

This is well defined because of associativity of vector addition.

## Definition

If  $\mathbf{a}_1, \dots, \mathbf{a}_n$  are the columns of  $A$  then

$$A\mathbf{x} := [\mathbf{a}_1 \quad \mathbf{a}_2 \dots \quad \mathbf{a}_n] \begin{bmatrix} x_1 \\ \cdot \\ \cdot \\ \cdot \\ x_n \end{bmatrix} = x_1\mathbf{a}_1 + \dots + x_n\mathbf{a}_n$$

The above definition agrees with the definition of  $A\mathbf{x}$ , when  $\mathbf{x}$  is viewed as a column matrix, as the  $k$ , 1-th entry of the column matrix  $A\mathbf{x}$  is the same as the  $k$ -th entry of  $A\mathbf{x}$ , viewed as a vector in  $\mathbb{R}^m$ .

## Theorem

If  $A$  is an  $m \times n$  matrix, with columns  $\mathbf{a}_1, \dots, \mathbf{a}_n$ , and if  $\mathbf{b}$  is in  $\mathbb{R}^m$ , the matrix equation

$$A\mathbf{x} = \mathbf{b}$$

has the same solution set as the vector equation

$$x_1\mathbf{a}_1 + \dots + x_n\mathbf{a}_n = \mathbf{b}$$

which, in turn, has the same solution set as the system of linear equations whose augmented matrix is

$$[\mathbf{a}_1 \quad \mathbf{a}_2 \quad \dots \quad \mathbf{a}_n \quad \mathbf{b}]$$

# Homogeneous Linear Systems

## Definition

A system of linear equations is said to be *homogeneous* if it can be written in the form

$$A\mathbf{x} = \mathbf{0},$$

where  $A$  is an  $m \times n$  matrix.

## Fact

The homogeneous equation  $A\mathbf{x} = \mathbf{0}$  has a nontrivial solution if and only if the equation has at least one free variable.

# Parametric Vector Form

Whenever the solution set of a linear system is described explicitly in terms of a linear combination of vectors with variable and/or fixed weights, we say that the solution is in *parametric vector form*.

For a homogeneous system, the weights are all variable.

For the non-homogeneous system

$$A\mathbf{x} = \mathbf{b}, \quad \mathbf{b} \neq \mathbf{0}$$

we first check if the system is consistent.



## Writing a Solution Set (of a consistent system) in Parametric Vector Form

- 1 Row reduce the augmented matrix to reduced echelon form.
- 2 Express each basic variable in terms of any free variables appearing in an equation.
- 3 Write a typical solution  $\mathbf{x}$  as a vector whose entries depend on the free variables, if any.
- 4 Decompose  $\mathbf{x}$  into a linear combination of vectors (with numeric entries) using the free variables as parameters.

$$x_1 - 3x_2 - 5x_3 = 0$$

$$x_2 + x_3 = 3$$

### Example

Find the general solution of the linear system whose augmented matrix is

$$\left[ \begin{array}{ccc|c} 1 & -3 & -5 & 0 \\ 0 & 1 & 1 & 3 \end{array} \right]$$

Let us reduce this to RREF.

$$\xrightarrow{R_1 \rightarrow R_1 + 3R_2} \left[ \begin{array}{ccc|c} 1 & 0 & -2 & 9 \\ 0 & 1 & 1 & 3 \end{array} \right]$$

Let us write the system of equations.

$$x_1 - 2x_3 = 9$$

$$x_2 + x_3 = 3$$

Rewrite expressing basic variables in terms of free variables (if any).

$$\begin{aligned}x_1 &= 9 + 2x_3 \\x_2 &= 3 - x_3 \\x_3 &\text{ free}\end{aligned}$$

In Parametric Vector Form:

$$\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \mathbf{x} = \begin{bmatrix} 9 \\ 3 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} 2 \\ -1 \\ 1 \end{bmatrix}$$

$$\text{The solution set of } Ax = b \\ = \{ x \in \mathbb{R}^n \mid Ax = b \} = S$$

## Theorem

Suppose the equation

$$Ax = b$$

is consistent for some given  $b$ , and let  $p$  be a solution (i.e.  $Ap = b$ ). Then the solution set of  $Ax = b$  is the set of all vectors of the form

$$w = p + v_h,$$

where  $v_h$  is any solution of the homogeneous equation

$$\underline{Ax = 0}.$$

$$T = \{ w \mid w = p + v_h, \text{ where } v_h \in \mathbb{R}^n \text{ and } Av_h = 0 \}$$

all of  
① Make sure your symbols  
are well-defined when  
you write a proof.

# Proof

Suppose  $\mathbf{w}$  is any solution of the equation  $A\mathbf{x} = \mathbf{b}$ . This means

$$A\mathbf{w} = \mathbf{b}. \quad (1)$$

Also, we're given that

$$A\mathbf{p} = \mathbf{b}. \quad (2)$$

If we subtract equation (2) from (1) we get

$$A(\mathbf{w} - \mathbf{p}) = \mathbf{0}.$$

Hence  $\mathbf{w} - \mathbf{p}$  is a solution of the equation  $A\mathbf{x} = \mathbf{0}$ .

Put  $\mathbf{v}_h = \mathbf{w} - \mathbf{p}$ . Then

$$\underline{\mathbf{w} = \mathbf{p} + \mathbf{v}_h} = \mathbf{p} + (\mathbf{w} - \mathbf{p})$$

Next we show that any vector of the form  $p + v_h$  is a solution of  $Ax = b$ , where  $p$  is a particular solution and  $v_h$  is any solution of the homogeneous system  $Ax = 0$ . This means

$$\underline{Av_h = 0.} \quad (3)$$

Put  $\underline{w = p + v_h}$ . Then using equations (2) and (3) we get


$$\underline{Aw = Ap + Av_h = b.}$$

which means  $w$  is a solution of  $Ax = b$ .

$$\underline{Aw = A(p + v_h) = Ap + Av_h = Ap + 0 = b.}$$

To show that two sets  $S$   
and  $T$  are equal, we have

to show  $S \subset T$  as well

as  $T \subset S$ .  subset relation



# Matrices - Review

## Theorem

Let  $A$ ,  $B$ , and  $C$  be  $m \times n$  matrices, and let  $r$  and  $s$  be scalars.

- a.  $A + B = B + A$  (Commutativity of addition)
- b.  $A + (B + C) = (A + B) + C$  (Associativity of addition)
- c.  $A + 0 = A$  (Additive Identity)
- d.  $r(A + B) = rA + rB$  (Distributive Law)
- e.  $(r + s)A = rA + sA$  (Distributive Law)
- f.  $r(sA) = (rs)A$

# Matrix Multiplication

Let  $A = (a_{ij})$  be an  $m \times n$  matrix and  $B = (b_{ij})$  be an  $n \times p$  matrix. We define the matrix product  $C := AB$  as the  $m \times p$  matrix whose entries are

$$c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}. \quad (\text{Row-Column Rule})$$

## Theorem

Let  $A$  be an  $m \times n$  matrix, and let  $B$  and  $C$  have sizes for which the indicated sums and products are defined.

- a.  $A(BC) = (AB)C$  (associative law of multiplication)
- b.  $A(B + C) = AB + AC$  (left distributive law)
- c.  $(B + C)A = BA + CA$  (right distributive law)
- d.  $r(AB) = (rA)B = A(rB)$  for any scalar  $r$
- e.  $I_m A = A = A I_n$  (identity for matrix multiplication)

Let's verify the associative law.

## Powers of a Matrix

$$A^k = \underbrace{A \dots A}_k$$

## Transpose of a Matrix

Let  $A$  and  $B$  denote matrices whose sizes are appropriate for the following sums and products.

- a.  $(A^T)^T = A$
- b.  $(A + B)^T = A^T + B^T$
- c.  $(rA)^T = rA^T$  for any scalar  $r$
- d.  $(AB)^T = B^T A^T$

## A New Perspective

$$B = [b_1 \ \dots \ b_p]$$

This matrix product has the property that if  $\mathbf{b}_1, \dots, \mathbf{b}_p$  are the columns of  $B$  then

$$AB = [\underbrace{A\mathbf{b}_1} \quad A\mathbf{b}_2 \quad \dots \quad A\mathbf{b}_p]$$

Each column of  $AB$  is a linear combination of the columns of  $A$  using weights from the corresponding column of  $B$ .

$3 \times 2$

$a_1$   $a_2$

$A =$

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

$B =$

$$\begin{bmatrix} 5 & 1 & 3 \\ 1 & 2 & 0 \end{bmatrix}$$

$2 \times 3$

$b_1$   $b_2$   $b_3$

$$AB = \begin{bmatrix} 6 & 3 & 3 \\ 10 & 2 & 6 \\ 16 & 5 & 9 \end{bmatrix}$$

$=$

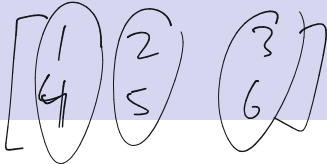
$$\begin{bmatrix} Ab_1 & Ab_2 & Ab_3 \end{bmatrix}$$

Row col rule:

$$AB = \begin{bmatrix} 6 & 3 & 3 \\ 10 & 2 & 6 \\ 16 & 5 & 9 \end{bmatrix}$$

$$Ab_1 = 5 \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} + 1 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$$

# Proof



Let  $A = (a_{ij})$  be an  $m \times n$  matrix and let  $B = (b_{ij})$  be an  $n \times p$  matrix whose columns are  $\mathbf{b}_1, \dots, \mathbf{b}_p$ .

Let  $k \in \{1, \dots, p\}$ . Then for  $1 \leq l \leq m$ , the  $l$ -th entry of the  $k$ -th column of  $AB$  is simply the  $l, k$ -th entry of the matrix product  $AB$ , which is

$$\sum_{j=1}^n a_{lj} b_{jk}$$

If we denote the columns of  $A$  by  $\mathbf{a}_1, \dots, \mathbf{a}_n$ , then

$$\underline{A\mathbf{b}_k} = \underline{b_{1k}\mathbf{a}_1 + \dots + b_{nk}\mathbf{a}_n} = \underbrace{\sum_{j=1}^n b_{jk}\mathbf{a}_j}$$

$$\mathbf{b}_k = \begin{bmatrix} b_{1k} \\ b_{2k} \\ \vdots \\ b_{nk} \end{bmatrix}$$

$$b_{jr} a_j = \begin{bmatrix} b_{jr} a_{1j} \\ b_{jr} a_{2j} \\ \boxed{b_{jr} a_{lj}} \\ b_{jr} a_{mj} \end{bmatrix} \quad \text{--- } l\text{-th entry}$$

Thus the  $l$ -th entry of the column vector  $A\mathbf{b}_k$  is

$$\sum_{j=1}^n \underline{b_{jk} a_{lj}} = \sum_{j=1}^n a_{lj} b_{jk}$$

Therefore  $A\mathbf{b}_k$  is the  $k$ -th column of  $AB$ .

Similarly, each row of  $\underline{AB}$  is a linear combination of the rows of  $\underline{B}$  using weights from the corresponding row of  $A$ . In other words

$$\underline{row_i(AB)} = \underline{row_i(A)}B.$$

**Proof.**

$$\begin{aligned} row_i(AB) &= (col_i((AB)^T))^T \\ &= (col_i(B^T A^T))^T \\ &= (B^T col_i(A^T))^T \\ &= row_i(A)B. \end{aligned}$$





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

$$B = \begin{bmatrix} 5 & 1 & 3 \\ 1 & 2 & 0 \end{bmatrix}$$

$$\text{row}_1(AB) = 1 \begin{bmatrix} 5 & 1 & 3 \end{bmatrix} - 1 \begin{bmatrix} 1 & 2 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 6 & 3 & 3 \end{bmatrix}$$

$$\begin{aligned} \text{row}_2(AB) &= 2 \begin{bmatrix} 5 & 1 & 3 \end{bmatrix} + 0 \begin{bmatrix} 1 & 2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 10 & 2 & 6 \end{bmatrix} \end{aligned}$$

$$\text{row}_3(AB) = 3 \begin{bmatrix} 5 & 13 \end{bmatrix} + 1 \begin{bmatrix} 1 & 20 \end{bmatrix}$$

$$= \begin{bmatrix} 16 & 59 \end{bmatrix}$$

$$AB = \begin{bmatrix} 6 & 3 & 3 \\ 10 & 2 & 6 \\ 16 & 5 & 9 \end{bmatrix}$$

Inconsistent system with  
free variable.

$$x_1 + x_2 = 1$$

$$x_1 + x_2 = 2$$

not pivot  
column

$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$$

$$5x_4 = 0$$