

Section 1.4: The Matrix Equation $A\mathbf{x} = \mathbf{b}$.

Definition Let A be an $m \times n$ matrix whose columns are the vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ (each in \mathbb{R}^m), and let \mathbf{x} be a vector in \mathbb{R}^n . Then the product of A and \mathbf{x} , denoted by

$$A\mathbf{x}$$

is the linear combination of the columns of A whose weights are the corresponding entries in \mathbf{x} . That is

$$A\mathbf{x} = x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n.$$

(Note that the result is a vector in \mathbb{R}^m !)

Theorem

If A is the $m \times n$ matrix whose columns are the vectors $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$, and \mathbf{b} is in \mathbb{R}^m , then the matrix equation

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

has the same solution set as the vector equation

$$x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n = \mathbf{b}$$

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

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

Corollary

The equation $A\mathbf{x} = \mathbf{b}$ has a solution if and only if \mathbf{b} is a linear combination of the columns of A .

In other words, the corresponding linear system is consistent if and only if \mathbf{b} is in $\text{Span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}$.

Example

Characterize the set of all vectors $\mathbf{b} = (b_1, b_2, b_3)$ such that $A\mathbf{x} = \mathbf{b}$ has a solution where

$$A = \begin{bmatrix} 1 & 3 & 4 \\ -4 & 2 & -6 \\ -3 & -2 & -7 \end{bmatrix}.$$

Last time, we did row reduction on the matrix $[A \ \mathbf{b}]$ and found that the equation is solvable only for vectors \mathbf{b} such that

$$-2b_1 + b_2 - 2b_3 = 0. \quad \Rightarrow \quad b_1 = \frac{1}{2}b_2 - b_3$$

We can state this result in terms of a subset of \mathbb{R}^3 .

We can restate the set of vectors \vec{b} as

$$\vec{b} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}b_2 - b_3 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}b_2 \\ b_2 \\ 0 \end{bmatrix} + \begin{bmatrix} -b_3 \\ 0 \\ b_3 \end{bmatrix}$$

$$= b_2 \begin{bmatrix} \frac{1}{2} \\ 1 \\ 0 \end{bmatrix} + b_3 \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$$

b_2, b_3 are any
real numbers

So our allowable vectors \vec{b} are linear combinations of the vectors $\begin{bmatrix} \frac{1}{2} \\ 1 \\ 0 \end{bmatrix}$ and $\begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix}$.

We can say that $A\vec{x} = \vec{b}$ is consistent if

\vec{b} is in $\text{Span} \left\{ \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 1 \end{bmatrix} \right\}$.

Theorem (first in a string of equivalency theorems)

Let A be an $m \times n$ matrix. Then the following are logically equivalent (i.e. they are either all true or are all false).

- (a) For each \mathbf{b} in \mathbb{R}^m , the equation $A\mathbf{x} = \mathbf{b}$ has a solution.
- (b) Each \mathbf{b} in \mathbb{R}^m is a linear combination of the columns of A .
- (c) The columns of A span \mathbb{R}^m .
- (d) A has a pivot position in every row.

(Note that statement (d) is about the *coefficient* matrix A , not about an augmented matrix $[A \ \mathbf{b}]$.)

A Scalar Product

If \mathbf{u} and \mathbf{v} are vectors in \mathbb{R}^n , we define a scalar product (also called the *dot product*) via

$$\mathbf{u} \cdot \mathbf{v} = u_1 v_1 + u_2 v_2 + \cdots + u_n v_n$$

Example Compute $\mathbf{u} \cdot \mathbf{v}$ if $\mathbf{u} = (1, 2, 3)$ and $\mathbf{v} = (-1, 0, 4)$.

$$\vec{u} \cdot \vec{v} = 1(-1) + 2(0) + 3(4) = 11$$

Computing Ax

We can use a *row-vector* dot product rule. The i^{th} entry in Ax is the sum of products of corresponding entries from row i of A with those of x . For example

$$\begin{bmatrix} 1 & 0 & -3 \\ -2 & -1 & 4 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix} = \begin{bmatrix} 1(2) + 0(1) + (-3)(-1) \\ -2(2) + (-1)(1) + 4(-1) \end{bmatrix} = \begin{bmatrix} 5 \\ -9 \end{bmatrix}$$

2×3

\mathbb{R}^3

product
in \mathbb{R}^2

1st entry: Dot product of row 1 of A
with \vec{x}

2nd entry: Dot product of row 2 of
 A with \vec{x} .

$$\begin{bmatrix} 2 & 4 \\ -1 & 1 \\ 0 & 3 \end{bmatrix} \begin{bmatrix} -3 \\ 2 \end{bmatrix} = \begin{bmatrix} 2(-3) + 4(2) \\ -1(-3) + 1(2) \\ 0(-3) + 3(2) \end{bmatrix} = \begin{bmatrix} 2 \\ 5 \\ 6 \end{bmatrix}$$

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \cdot x_1 + 0 \cdot x_2 + 0 \cdot x_3 \\ 0 \cdot x_1 + 1 \cdot x_2 + 0 \cdot x_3 \\ 0 \cdot x_1 + 0 \cdot x_2 + 1 \cdot x_3 \end{bmatrix} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$$

Identity Matrix

We'll call an $n \times n$ matrix with 1's on the diagonal and 0's everywhere else—i.e. one that looks like

$$\begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$$

the $n \times n$ **identity** matrix and denote it by I_n . (We'll drop the subscript if it's obvious from the context.)

This matrix has the property that for each \mathbf{x} in \mathbb{R}^n

$$I_n \mathbf{x} = \mathbf{x}.$$

Theorem: Properties of the Matrix Product

If A is an $m \times n$ matrix, \mathbf{u} and \mathbf{v} are vectors in \mathbb{R}^n , and c is any scalar, then

(a) $A(\mathbf{u} + \mathbf{v}) = A\mathbf{u} + A\mathbf{v}$, and

(b) $A(c\mathbf{u}) = cA\mathbf{u}$.

Section 1.5: Solution Sets of Linear Systems

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

$$Ax = \mathbf{0}$$

for some $m \times n$ matrix A and where $\mathbf{0}$ is the zero vector in \mathbb{R}^m .

Theorem: A homogeneous system $Ax = \mathbf{0}$ always has at least one solution $\mathbf{x} = \mathbf{0}$.

The solution $\mathbf{x} = \mathbf{0}$ is called the **trivial solution**. A more interesting question for a homogeneous system is

Does it have a nontrivial solution?

Theorem

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

Example: Determine if the homogeneous system has a nontrivial solution. Describe the solution set.

$$(a) \quad \begin{aligned} 2x_1 + x_2 &= 0 \\ x_1 - 3x_2 &= 0 \end{aligned}$$

we can state this as $A\vec{x} = \vec{0}$

where $A = \begin{bmatrix} 2 & 1 \\ 1 & -3 \end{bmatrix}$, $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

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

This has augmented matrix

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

ref \rightarrow

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

There are no free variables, hence there are no nontrivial solutions.

$$\begin{aligned} \text{The eqns read as } & x_1 = 0 \\ & x_2 = 0 \end{aligned}$$

The only solution is $\vec{x} = \vec{0}$.

Since the system is homogeneous, we can do rref on the coefficient matrix.

$$\begin{bmatrix} 2 & 1 \\ 1 & -3 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\begin{aligned} (b) \quad & 3x_1 + 5x_2 - 4x_3 = 0 \\ & -3x_1 - 2x_2 + 4x_3 = 0 \\ & 6x_1 + x_2 - 8x_3 = 0 \end{aligned}$$

An augmented matrix is

$$\begin{bmatrix} 3 & 5 & -4 & 0 \\ -3 & -2 & 4 & 0 \\ 6 & 1 & -8 & 0 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & -4/3 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

↑
non pivot column

With a free variable, the system has nontrivial solutions.

From the rref

$$x_1 = \frac{4}{3}x_3$$

$$x_2 = 0$$

x_3 - free

All non trivial solutions have the form

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \frac{4}{3}x_3 \\ 0 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 4/3 \\ 0 \\ 1 \end{bmatrix}$$

In fact, all solutions are of the

form

$$\vec{x} = x_3 \begin{bmatrix} 4/3 \\ 0 \\ 1 \end{bmatrix}, \quad x_3 \text{ any real number.}$$

(c) $x_1 - 2x_2 + 5x_3 = 0$

This has coefficient matrix $[1 \ -2 \ 5]$

It's already in rref. There are free variables, hence there are nontrivial solutions. \vec{x} has to satisfy

$$x_1 = 2x_2 - 5x_3$$

x_2, x_3 - free

As a vector

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2x_2 - 5x_3 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 2x_2 \\ x_2 \\ 0 \end{bmatrix} + \begin{bmatrix} -5x_3 \\ 0 \\ x_3 \end{bmatrix}$$

$$= x_2 \begin{bmatrix} 2 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -5 \\ 0 \\ 1 \end{bmatrix} \quad x_2, x_3 \text{ are free}$$

Parametric Vector Form of a Solution Set

Example (b) had a solution set consisting of vectors of the form $\mathbf{x} = x_3\mathbf{v}$. Example (c)'s solution set consisted of vector that look like $\mathbf{x} = x_2\mathbf{u} + x_3\mathbf{v}$. Since these are **linear combinations**, we could write the solution sets like

$$\text{Span}\{\mathbf{u}\} \quad \text{or} \quad \text{Span}\{\mathbf{u}, \mathbf{v}\}.$$

Instead of using the variables x_2 and/or x_3 we often substitute **parameters** such as s or t .

The forms

$$\mathbf{x} = s\mathbf{u}, \quad \text{or} \quad \mathbf{x} = s\mathbf{u} + t\mathbf{v}$$

are called **parametric vector forms**.