

Lecture 3: Matrix Equations

Multiplying a matrix by a (column) vector: 2 approaches

① Row-by-column (more computationally efficient)

row vectors

$$\begin{pmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} \stackrel{\text{definition}}{=} \begin{pmatrix} a_1x + b_1y + c_1z \\ a_2x + b_2y + c_2z \end{pmatrix}$$

take "dot products"

Notice that this

$$= \begin{pmatrix} a_1x \\ a_2x \end{pmatrix} + \begin{pmatrix} b_1y \\ b_2y \end{pmatrix} + \begin{pmatrix} c_1z \\ c_2z \end{pmatrix}$$
$$= x \begin{pmatrix} a_1 \\ a_2 \end{pmatrix} + y \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} + z \begin{pmatrix} c_1 \\ c_2 \end{pmatrix},$$

which leads to ...

② Linear combinations (more conceptual)

Writing $A = \begin{pmatrix} \uparrow & \uparrow & \dots & \uparrow \\ \vec{v}_1 & \vec{v}_2 & \dots & \vec{v}_n \\ \downarrow & \downarrow & \dots & \downarrow \end{pmatrix}$, $\vec{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$,

column vectors

(*)

$$A \vec{x} = x_1 \vec{v}_1 + x_2 \vec{v}_2 + \dots + x_n \vec{v}_n$$

As an example of the second approach's utility, we can use it to check the

Linearity property: $A(c\vec{u} + d\vec{w}) = cA\vec{u} + dA\vec{w}$

$\underbrace{\begin{pmatrix} cu_1 + dw_1 \\ \vdots \\ cu_n + dw_n \end{pmatrix}}_{\text{components}} \quad \underbrace{\begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}}_{\text{vector } \vec{u}} \quad \underbrace{\begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix}}_{\text{vector } \vec{w}}$

By (*),

$$\begin{aligned} \text{LHS} &= (cu_1 + dw_1)\vec{v}_1 + \dots + (cu_n + dw_n)\vec{v}_n \\ &= c(u_1\vec{v}_1 + \dots + u_n\vec{v}_n) + d(w_1\vec{v}_1 + \dots + w_n\vec{v}_n) \\ &= \text{RHS}. \end{aligned}$$

More importantly, we see that

Statement 1: $A\vec{x} = \vec{b}$ has a solution in \vec{x}
(i.e. is consistent)

↕ is equivalent to

Statement 2: $\vec{b} \in \text{Span}\{\text{columns of } A\}$

The link is (*): $\vec{b} = A\vec{x} \stackrel{(*)}{=} x_1\vec{v}_1 + x_2\vec{v}_2 + \dots + x_n\vec{v}_n$
says "you can choose x_1, \dots, x_n so that \vec{b} is a linear combination of $\vec{v}_1, \dots, \vec{v}_n$ ".

So to check Statement 2, you just row-reduce $[A | \vec{b}]$.

there is another set of equivalent statements:

Assertion (A): $\text{Span}\{\text{columns of } A\} = (\text{all of}) \mathbb{R}^m$

\Updownarrow (clear from above)

Assertion (B): $A\vec{x} = \vec{b}$ is consistent for any \vec{b}

\Updownarrow why?

Assertion (C): $\text{rref}(A)$ has no rows of all zeroes.

First,

• $[A \mid \vec{b}] \underset{\text{row-equiv.}}{\sim} [\text{rref}(A) \mid \vec{c}]$ for some vector \vec{c}

(apply the sequence of row operations that puts A in RREF, to the augmented matrix)

• we can choose \vec{b} so that \vec{c} is any vector (because row operations are reversible)

(C) \Rightarrow (B): If (C) holds, then $[\text{rref}(A) \mid \vec{c}]$

has a leading '1' in every row in the " $\text{rref}(A)$ " part, hence is in RREF itself and $= \text{rref}[A \mid \vec{b}]$. Since the

leading '1's occur to the left of \vec{c} , \vec{c} is not a

pivot column and the system is consistent (regardless of \vec{b}). So (B) holds.

(B) \Rightarrow (C): If (C) fails, choose \vec{b} so that \vec{c} has a nonzero last entry. Since the last row of $\text{ref}(A)$ is all '0's, \vec{c} is a pivot column (for this choice of \vec{b}), and so (B) fails.

Ex 1 / For which \vec{b} is $\begin{pmatrix} 3 & -1 \\ -9 & 3 \end{pmatrix} \vec{x} = \vec{b}$ solvable?

(Equivalent question: determine $\text{span}\left\{\begin{pmatrix} 3 \\ -9 \end{pmatrix}, \begin{pmatrix} -1 \\ 3 \end{pmatrix}\right\}$.)

$$\left[\begin{array}{cc|c} 3 & -1 & b_1 \\ -9 & 3 & b_2 \end{array} \right] \rightarrow \left[\begin{array}{cc|c} 1 & -\frac{1}{3} & b_1/3 \\ -9 & 3 & b_2 \end{array} \right] \rightarrow \left[\begin{array}{cc|c} 1 & -\frac{1}{3} & b_1/3 \\ 0 & 0 & b_2 + 3b_1 \end{array} \right]$$

so we must have $b_2 = -3b_1$, i.e. \vec{b} must be a scalar multiple of the vector $\begin{pmatrix} 1 \\ -3 \end{pmatrix}$.

failure of (C)

failure of (A)

Ex 2 / Do the columns of $A = \begin{pmatrix} 1 & 2 & 3 & 4 \\ 5 & 6 & 7 & 8 \\ 9 & 10 & 11 & 12 \end{pmatrix}$ span \mathbb{R}^3 ?

(Equivalent question: does $\text{ref}(A)$ have no rows of "all 0"?)

Row-reduce:

$$A \rightarrow \begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & -4 & -8 & -12 \\ 0 & -8 & -16 & -24 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 2 & 3 & 4 \\ 0 & 1 & 2 & 3 \\ 0 & -8 & -16 & -24 \end{pmatrix} \rightarrow \begin{pmatrix} 1 & 0 & -1 & -2 \\ 0 & 1 & 2 & 3 \\ 0 & 0 & 0 & 0 \end{pmatrix} = \text{ref}(A)$$

Answer: NO.

Ex 3 / Suppose $A = 4 \times 4$ matrix, and $\vec{b} \in \mathbb{R}^4$ are such that $A\vec{x} = \vec{b}$ has a unique solution. Must columns of A span \mathbb{R}^4 ?

Consider the augmented matrix $[A \mid \vec{b}]$ and its rref, which must have

- \vec{b} non-pivot column (for existence of a solution)
- no non-pivot columns in the "[...]" part of the augmented matrix (for uniqueness)
= free variables

So all columns of A are pivot } \Rightarrow all rows of rref have a leading '1'
AND A 4×4

\Rightarrow no rows of zeros

\Rightarrow columns span \mathbb{R}^4 .

Notice that if A was instead 5×4 , this argument breaks down and columns need not span \mathbb{R}^5 (in fact, they can't).

In lecture 2, I claimed the

Theorem: Every matrix A is row-equivalent to a unique RREF matrix.

To show this, we need the following:

(*) If A is row-equivalent to B , then the rows of A are linear combinations of rows of B (and vice versa).

This is because

- the row operations — replace, swap, scale — simply replace a given row by a linear combination of rows.
- they are reversible.

To prove the Theorem, proceed in 3 steps:

① It is enough to show that

(**) 2 row-equivalent RREF matrices U and V must be the same.

Why? By the algorithm, $A \underset{\text{row-eq.}}{\sim} \text{rref}(A) =: U$.

If $A \underset{\text{row-eq.}}{\sim} V$ (also RREF), then $U \underset{\text{row-eq.}}{\sim} V$.

(So if (***) is known, we get $U = V$ as desired.)

For the remaining 2 steps, (which prove (**)), let U and V be 2 row-equivalent RREF matrices:

② The pivot columns of U & V are the same.

picture of U

$$\begin{bmatrix} 0 \dots 0 & | & * \dots * & 0 & * \dots * & 0 \dots \\ 0 \dots \dots \dots & & & 0 & | & * \dots * & 0 \dots \\ 0 \dots \dots \dots & & & & & & 0 & | \dots \\ & & & & & & & \vdots \\ & & & & & & & \vdots \end{bmatrix}$$

column: $i_1 \quad i_2 \quad i_3 \dots$

picture of V

$$\begin{bmatrix} 0 \dots 0 & | & * \dots * & 0 & * \dots * & 0 \dots \\ 0 \dots \dots \dots & & & 0 & | & * \dots * & 0 \dots \\ 0 \dots \dots \dots & & & & & & 0 & | \dots \\ & & & & & & & \vdots \\ & & & & & & & \vdots \end{bmatrix}$$

column: $j_1 \quad j_2 \quad j_3 \dots$

Write \vec{u}_1, \vec{u}_2 , etc. for rows of U (viewed as vectors), same for V .

(A) By (x), $\vec{u}_1 \in \text{span}\{\text{rows of } V\}$. So:

$$\vec{u}_1 = a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots \Rightarrow i_1 \geq j_1.$$

Reversing U & V gives $j_1 \geq i_1$. So $i_1 = j_1$ (and $a_1 = 1$).

(B) Next, $\vec{u}_2 = b_1 \vec{v}_1 + b_2 \vec{v}_2 + \dots \Rightarrow \vec{u}_2$ has b_1 in $(j_1 =) i_1^{\text{th}}$ coord.

$$\Rightarrow b_1 = 0$$

$$\Rightarrow \vec{u}_2 = b_2 \vec{v}_2 + b_3 \vec{v}_3 + \dots$$

(and vice versa).

(C) Striking out the first row of U and V , the remaining (still RREF!) matrices are row equivalent. Go back to (A), find $i_2 = j_2$, and repeat until there's nothing left.

③ The non-pivot columns of U & V are equal.

(D) We have $\vec{u}_1 = \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_m \vec{v}_m$ (recall $a_1 = 1$ above)

$\Rightarrow \vec{u}_1$ has a_2 in the $(j_2 =) i_2^{\text{th}}$ coordinate

a_3 in the $(j_3 =) i_3^{\text{th}}$ coordinate, etc

$$\Rightarrow 0 = a_2 = a_3 = \dots = a_m$$

So $\vec{u}_1 = \vec{v}_1$.

(E) Stride out the first row of U & V , go back to (D),
get $\vec{u}_2 = \vec{v}_2$, and so on. □