

5.

Rows vs. columns

Note: everything today works over any field.

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$$A \in \mathbb{R}^{m \times n} \Rightarrow A = \begin{bmatrix} -A_1- \\ -A_i- \\ \vdots \\ -A_m- \end{bmatrix} \quad A_i = [a_{i1} \dots a_{in}] \in \mathbb{R}_{\text{row}}^n \quad i = \text{generic row index}$$

$$= \begin{bmatrix} | & | & | & | \\ a_1 & \dots & a_j & \dots & a_n \\ | & | & | & | \end{bmatrix} \quad a_j = \begin{bmatrix} a_{1j} \\ \vdots \\ a_{mj} \end{bmatrix} \in \mathbb{R}_{\text{col}}^m \quad j = \text{generic column index}$$

$$Ax = b \Leftrightarrow A_i x = b_i \text{ for } i = 1, \dots, m$$

but also

$$Ax = b \Leftrightarrow x_1 a_1 + \dots + x_n a_n = b$$

$\Leftrightarrow Ax$ is a linear combin of the a_j
 $\Leftrightarrow b$ is a linear combin of the columns of A
 $\Leftrightarrow b$ lies in span of cols of A

$$\Leftrightarrow b \in \text{span}(a_1, \dots, a_n)$$

$$\begin{array}{c} x_1 a_{11} + \dots + x_n a_{1n} \\ \vdots \\ x_1 a_{m1} + \dots + x_n a_{mn} \end{array} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$$

Def: $C(A) = \underbrace{\text{column space}}$ of A

or image of M_A

why? $A \rightsquigarrow$ function taking x 's to b 's:

$$\text{E.g. } \begin{bmatrix} -3 & 6 & 4 & e \\ 1 & \frac{2}{3} & -1 & 0 \end{bmatrix} \begin{bmatrix} 3 \\ 2 \\ 4 \\ \pi \end{bmatrix} = 3 \begin{bmatrix} -3 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 6 \\ \frac{2}{3} \end{bmatrix} + 4 \begin{bmatrix} 4 \\ -1 \end{bmatrix} + \pi \begin{bmatrix} e \\ 0 \end{bmatrix}$$

have class fill these in

Q. When does $Ax = b$ have a solution?

i.e., When is $Ax = b$ consistent?

A. ✓ When there exists an x listing the coefficients in a linear combination of the columns of A that equals b .

✓ When b lies in span of columns of A

✓ When $b \in C(A)$

✓ When b lies in the image of the function "multiplication by A " M_A

$$\begin{array}{ccc} \mathbb{R}^m & \xleftarrow{\quad} & \mathbb{R}^n \\ \uparrow & & \uparrow \\ Ax & & x \end{array}$$

$$\begin{bmatrix} Ax \\ \vdots \\ A \end{bmatrix} \xleftarrow{\quad} \boxed{A} \xrightarrow{\quad} \begin{bmatrix} 1 \\ \vdots \\ x \end{bmatrix}$$

$$\begin{array}{ccc} \begin{bmatrix} -3 & 6 & 4 & e \\ 1 & \frac{2}{3} & -1 & 0 \end{bmatrix} & \xleftarrow{\quad} & \begin{bmatrix} 1 \\ \vdots \\ x \end{bmatrix} \\ \xrightarrow{\quad} & & \begin{bmatrix} 1 \\ \vdots \\ y_3 \end{bmatrix} \end{array}$$

$$\begin{array}{ccc} \begin{bmatrix} 1 \\ \vdots \\ y_3 \end{bmatrix} & \xleftarrow{\quad} & \begin{bmatrix} 3 \\ 2 \\ 4 \\ \pi \end{bmatrix} \\ \uparrow & & \uparrow \\ \text{"maps to"} & & \end{array}$$

In (reduced) echelon form:

$$\left[\begin{array}{cccc|c} 1 & 0 & 0 & 0 & b_1 \\ 0 & 1 & 0 & 0 & b_2 \\ 0 & 0 & 1 & 0 & b_3 \\ \vdots & \vdots & \vdots & \vdots & b_r \\ 0 & 0 & 0 & 1 & b_{r+1} \\ \hline 0 & 0 & 0 & 0 & b_m \end{array} \right]$$

Lemma 1: consistent $\Leftrightarrow b_{r+1} = 0$

$$\vdots \quad b_m = 0$$

0_x

Pf: $\Rightarrow A_i = 0$ for $i > r$ so $Ax = b \Rightarrow A_i x = b_i$ for $i > r$.

\Leftarrow : For any values of the free variables, the values of the pivots are determined from bottom to top.

E.g.

$$\left[\begin{array}{cccc|c} 1 & \pi & 4 & 1 & 7 \\ 0 & 0 & 1 & 3 & -\frac{11}{5} \\ 0 & 0 & 0 & 0 & ? \end{array} \right]$$

pivot pivot free free

$x_1 + \pi x_2 + 4x_3 + x_4 = 7$
 $x_3 + 3x_4 = -\frac{11}{5}$

$0 + 0 + 0 + 0 - ? = ?$

\Rightarrow make whatever you want!
 consistent if $? = 0$, inconsistent if $? \neq 0$

Def: The rank of $A \in \mathbb{R}^{m \times n}$ is $\dim C(A) = \min \# \text{columns required to span } C(A)$

E.g. rank $\left[\begin{array}{cccc} 1 & \pi & 4 & 1 \\ 0 & 0 & 1 & 3 \\ 0 & 0 & 0 & 0 \end{array} \right] = ?$ 2 rank $\left[\begin{array}{cccc|c} 1 & \pi & 4 & 1 & 7 \\ 0 & 0 & 1 & 3 & -\frac{11}{5} \\ 0 & 0 & 0 & 0 & ? \end{array} \right] = ?$ 2 if $? = 0$
 3 if $? \neq 0$

This is the "correct" definition that I mentioned on day 1. Learn it. Use it.
 Reduce all questions about rank to it. You will be tested on it. It should be your go-to def.

Lemma 2: Fix v_1, \dots, v_r cols of A with corresponding cols

$$v'_1, \dots, v'_r \quad A', \text{ where } [A|b] \xrightarrow{\text{row op}} [A'|b'].$$

$$\text{Cor: } C(A) = \mathbb{R}^m \Rightarrow C(A') = \mathbb{R}^m.$$

$$\text{Pf: } b' \in \mathbb{R}^m \Rightarrow [A'|b'] \xrightarrow{\text{row op}} [A|b] \quad \text{for some } b \in C(A)$$

$$\text{Then } b = c_1 v_1 + \dots + c_r v_r \Leftrightarrow b' = c_1 v'_1 + \dots + c_r v'_r. \quad \text{Note: same coeffs } c_i. \quad \mathbb{R}^m. \square$$

Pf: Check type (i), (ii), (iii) row ops directly. And observe that $[A|b] \xrightarrow{\text{row op}} [A'|b'] \Leftrightarrow [A'|b'] \xrightarrow{\text{row op}} [A|b]. \quad \square$

Prop 1: v_1, \dots, v_r span $C(A) \Leftrightarrow v'_1, \dots, v'_r$ span $C(A')$ if $A \xrightarrow{\text{row op}} A'$.

Pf: $\Leftarrow b \in C(A) \Rightarrow b' \in C(A)$ by Lemma 2

$$\Rightarrow b' = c_1 v'_1 + \dots + c_r v'_r \text{ by hypothesis}$$

$$\Rightarrow b = c_1 v_1 + \dots + c_r v_r \text{ by Lemma.}$$

\Rightarrow same (i.e., swap the primes and un-primes) \square

This is why pivot columns are important.

Corollary: $\text{rank } A = \text{rank } A'$ if $A \xrightarrow{\text{row op}} A'$. Why is this useful?

Prop 2: $\text{rank } A = \# \text{ pivots in any echelon form, and pivot cols of } A \text{ minimally span } C(A)$.

Pf: By Cor and Prop 1, need only prove for $A = U$, reduced echelon form.

Lemma 1 \Rightarrow pivot cols $\begin{matrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \\ \vdots & \ddots & \vdots \\ 0 & 0 & 0 \\ 0 & 0 & 1 \end{matrix}$ span $C(U)$, and \mathbb{R}^r can't be spanned by less than r vectors: $C\left(\begin{matrix} < r \\ r \end{matrix}\right) = \mathbb{R}^r \Rightarrow$ reduced echelon form has column space \mathbb{R}^r by (Cor of) Lemma 2, but last row is 0, so that is impossible. \square

Prop 3: $Ax = b$ consistent $\Leftrightarrow \text{rank } A = \text{rank } [A|b]$. One direction is easy; which?

Pf: \Rightarrow : consistent $\Rightarrow b \in C(A) \Rightarrow C([A|b]) = C(A)$ and same pivot cols span minimally.

\Leftarrow : pivot cols of A span $C([A|b]) \Rightarrow b$ lies in their span. \square

Parametric to implicit implicit to parametric means: solve $Ax = b$.

Given x_0 and $v_1, \dots, v_k \in \mathbb{R}^n$, find linear equations $Ax = b$ so that $\text{sols}(Ax = b) = x_0 + \text{span}(v_1, \dots, v_k)$.

E.g. Find implicit equation(s) for the plane

$$x = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + s \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} + t \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix} \text{ in } \mathbb{R}^3.$$

Equivalently, which vectors x are expressible as

$$\text{Solve } x = \underbrace{\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}}_A \underbrace{\begin{bmatrix} s \\ t \end{bmatrix}}_x + \underbrace{\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}}_b, \text{ i.e.}$$

$$\underbrace{\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}}_A \underbrace{\begin{bmatrix} s \\ t \end{bmatrix}}_x = x - \underbrace{\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}}_b = \begin{bmatrix} x_1 - 1 \\ x_2 - 2 \\ x_3 - 1 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} + s \underbrace{\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}}_{\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}} + t \underbrace{\begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}}_{\begin{bmatrix} 1 & 2 \\ 0 & 1 \\ 1 & 1 \end{bmatrix}} ?$$

This is a linear system!

$$\text{cancel} \quad \left(\begin{array}{cc|c} 1 & 2 & x_1 - 1 \\ 0 & 1 & x_2 - 2 \\ 1 & 1 & x_3 - 1 \end{array} \right) \rightsquigarrow \left(\begin{array}{cc|c} 1 & 2 & x_1 - 1 \\ 0 & 1 & x_2 - 2 \\ 0 & -1 & x_3 - x_1 \end{array} \right)$$

$$\left(\begin{array}{cc|c} 1 & 2 & x_1 - 1 \\ 0 & 1 & x_2 - 2 \\ 0 & 0 & x_3 - x_1 + x_2 - 2 \end{array} \right)$$

has solution \Leftrightarrow every 0 row in echelon form has corresponding 0 on RHS,
 so x is expressible $\Leftrightarrow x_3 - x_1 + x_2 - 2 = 0$.