

**Example 101.** Let  $A$  be a  $3 \times 5$  matrix. For each of  $\text{col}(A)$ ,  $\text{row}(A)$  and  $\text{null}(A)$  determine  $d$  so that the space is a subspace of  $\mathbb{R}^d$ .

**Solution.**  $\text{col}(A)$  is a subspace of  $\mathbb{R}^3$ .  $\text{row}(A)$  is a subspace of  $\mathbb{R}^5$ .  $\text{null}(A)$  is a subspace of  $\mathbb{R}^5$ .

**Example 102.** Let  $A = \begin{bmatrix} 1 & 2 & 0 & 4 \\ 2 & 4 & 0 & 8 \\ 3 & 6 & 2 & 22 \\ 4 & 8 & 0 & 16 \end{bmatrix}$ . Find a basis for  $\text{col}(A)$ ,  $\text{row}(A)$  and  $\text{null}(A)$ .

**Solution.** We compute an echelon form of  $A$ :

$$\begin{bmatrix} 1 & 2 & 0 & 4 \\ 2 & 4 & 0 & 8 \\ 3 & 6 & 2 & 22 \\ 4 & 8 & 0 & 16 \end{bmatrix} \xrightarrow{\substack{R_2 - 2R_1 \Rightarrow R_2 \\ R_3 - 3R_1 \Rightarrow R_3 \\ R_4 - 4R_1 \Rightarrow R_4}} \begin{bmatrix} 1 & 2 & 0 & 4 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 10 \\ 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{R_2 \leftrightarrow R_3} \begin{bmatrix} 1 & 2 & 0 & 4 \\ 0 & 0 & 2 & 10 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

We can now use Theorem 99 to read off the bases we are interested in:

- The pivot columns are the first and third. Hence, a basis for  $\text{col}(A)$  is  $\begin{bmatrix} 1 \\ 2 \\ 3 \\ 4 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 2 \\ 0 \end{bmatrix}$ .

[Make sure you see why it would be horribly wrong to take columns from the echelon form.]

- $\begin{bmatrix} 1 \\ 2 \\ 0 \\ 4 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 2 \\ 10 \end{bmatrix}$  form a basis for  $\text{row}(A)$ .

[Note that it would be horribly wrong to take the first two rows from  $A$ .]

- The general solution to  $Ax = 0$  is  $x = \begin{bmatrix} -2s_1 - 4s_2 \\ s_1 \\ -5s_2 \\ s_2 \end{bmatrix} = s_1 \begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + s_2 \begin{bmatrix} -4 \\ 0 \\ -5 \\ 1 \end{bmatrix}$ .

Hence,  $\begin{bmatrix} -2 \\ 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -4 \\ 0 \\ -5 \\ 1 \end{bmatrix}$  are a basis for  $\text{null}(A)$ .

Let  $A$  be  $m \times n$ , and let  $r$  be the **rank** of  $A$ , that is,  $r$  is the number of pivots.

- $\dim \text{col}(A) = \dim \text{row}(A) = r$
- $\dim \text{null}(A) = n - r$

**Example 103.** The  $4 \times 4$  matrix  $A = \begin{bmatrix} 1 & 2 & 0 & 4 \\ 2 & 4 & 0 & 8 \\ 3 & 6 & 2 & 22 \\ 4 & 8 & 0 & 16 \end{bmatrix}$  from the previous example has rank 2.

Indeed,  $\dim \text{col}(A) = 2$ ,  $\dim \text{row}(A) = 2$ ,  $\dim \text{null}(A) = 4 - 2 = 2$  matches our computations.

**Example 104.** Let  $A$  be a  $3 \times 5$  matrix of rank 2. Determine the dimensions of  $\text{col}(A)$ ,  $\text{row}(A)$  and  $\text{null}(A)$ .

**Solution.**  $\dim \text{col}(A) = 2$ ,  $\dim \text{row}(A) = 2$ ,  $\dim \text{null}(A) = 5 - 2 = 3$ .

**Example 105.** Find a basis for  $\text{col}(A)$ ,  $\text{row}(A)$ ,  $\text{null}(A)$  with  $A = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 0 & 0 \\ 2 & 4 & 0 \end{bmatrix}$ .

**Solution.** For this simple matrix, we can just “see” the following (make sure you do, too!):

- A basis for  $\text{col}(A)$  is:  $\begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 4 \end{bmatrix}$  **Why?** Because these two vectors span and are clearly independent.

**Note.** We would select the same basis, if we computed an echelon form of  $A$  and applied Theorem 99(a).

- A basis for  $\text{row}(A)$  is:  $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ 0 \end{bmatrix}$  **Why?** Again, because these two vectors span and are clearly independent.

**Note.** An echelon form of  $A$  is  $\begin{bmatrix} 1 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 0 \end{bmatrix}$ . By Theorem 99(b), an alternative basis for  $\text{row}(A)$  is  $\begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 2 \\ 0 \end{bmatrix}$ .

In fact, further computing the RREF, we would select as basis  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}$  (probably the nicest basis for most purposes).

- A basis for  $\text{null}(A)$  is:  $\begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

**Why?** We know that  $\text{rank}(A) = 2$ . Hence,  $\dim \text{null}(A) = 3 - 2 = 1$ . Therefore, any nonzero vector in  $\text{null}(A)$  will be a basis for  $\text{null}(A)$ . Clearly,  $[0 \ 0 \ 1]^T$  is one such vector solving  $A\mathbf{x} = \mathbf{0}$  (why?).

### How little we actually know!

**Q:** How fast can we solve  $N$  linear equations in  $N$  unknowns?

Estimated cost of Gaussian elimination:

- |   |  |
|---|--|
| $\begin{bmatrix} \blacksquare & * & * & \cdots & * \\ 0 & * & * & \cdots & * \\ \vdots & \vdots & & & \vdots \\ 0 & * & * & \cdots & * \end{bmatrix}$ | <ul style="list-style-type: none"> <li>• to create the zeros below the first pivot:<br/><math>\implies</math> on the order of <math>N^2</math> operations</li> <li>• if there is <math>N</math> pivots total:<br/><math>\implies</math> on the order of <math>N \cdot N^2 = N^3</math> operations</li> </ul> |
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- A more careful count places the cost at  $\sim \frac{1}{3}N^3$  operations.

- For large  $N$ , it is only the  $N^3$  that matters.

It says that if  $N \rightarrow 10N$  then we have to work 1000 times as hard.

**That's not optimal!** We can do better than Gaussian elimination:

- Strassen algorithm (1969):  $N^{\log_2 7} = N^{2.807}$
- Coppersmith–Winograd algorithm (1990):  $N^{2.375}$
- ... Stothers–Williams–Le Gall (2014):  $N^{2.373}$  (If  $N \rightarrow 10N$  then we have to work 229 times as hard.)

Is  $N^{2+(\text{a tiny bit})}$  possible? **We don't know!** (People increasingly suspect so.) (Better than  $N^2$  is impossible; why?)

Good news for applications:

- Matrices typically have lots of structure and zeros which makes solving so much faster.