

## Bases for column and row spaces

**Example 112.** Find a basis for  $\text{col}(A)$  with  $A = \begin{bmatrix} 2 & 3 & 1 & 0 \\ 0 & 0 & 4 & 0 \end{bmatrix}$ .

**Solution.** Obviously,  $\text{col}(A) = \text{span}\left\{\begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \end{bmatrix}\right\} = \text{span}\left\{\begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \end{bmatrix}\right\}$ .

The vectors  $\begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \end{bmatrix}$  are also clearly independent, so they form a basis for  $\text{col}(A)$ .

**Solution.** We can also apply the recipe from Theorem 109. Since  $A$  is already in echelon form, we see directly that columns 2 and 4 correspond to free variables. Columns 1 and 3, that is  $\begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 4 \end{bmatrix}$ , are therefore selected as a basis.

**Example 113.** Find a basis for  $\text{col}(A)$  with  $A = \begin{bmatrix} 1 & 2 & -1 & 1 & 0 \\ 0 & 3 & 1 & -3 & 0 \\ 0 & 0 & 0 & 2 & 1 \end{bmatrix}$ .

**Solution.** Since  $A$  is already in echelon form, we can directly apply the recipe from Theorem 109.

Columns 3 and 5 correspond to a free variable, so a basis for  $\text{col}(A)$  is  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ -3 \\ 2 \end{bmatrix}$ .

[Note that many other choices would also lead to a basis: for instance, columns 1, 3, 4 also form a basis. Can you see that? Likewise, columns 2, 3, 5 form a basis as well.]

**Example 114.** Find a basis and the dimension of  $W = \text{span}\left\{\begin{bmatrix} 1 \\ 2 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}\right\}$ .

**Solution.** Is  $\dim W = 4$ ? No, because the third vector is the sum of the first two.

Suppose we did not notice...

$$\begin{bmatrix} 1 & 1 & 2 & 0 \\ 2 & 2 & 4 & 1 \\ 0 & 1 & 1 & 1 \\ 3 & 0 & 3 & 1 \end{bmatrix} \xrightarrow{\substack{R_2 - 2R_1 \Rightarrow R_2 \\ R_4 - 3R_1 \Rightarrow R_4}} \begin{bmatrix} 1 & 1 & 2 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 1 \\ 0 & -3 & -3 & 1 \end{bmatrix} \xrightarrow{\substack{\text{permute} \\ \text{rows}}} \begin{bmatrix} 1 & 1 & 2 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & -3 & -3 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{R_3 + 3R_2 \Rightarrow R_3} \begin{bmatrix} 1 & 1 & 2 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 4 \\ 0 & 0 & 0 & 1 \end{bmatrix} \xrightarrow{R_4 - \frac{1}{3}R_3 \Rightarrow R_4} \begin{bmatrix} 1 & 1 & 2 & 0 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

Not a pivot in every column, hence the 4 vectors are dependent.

Hence, a basis for  $W$  is  $\begin{bmatrix} 1 \\ 2 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1 \\ 1 \end{bmatrix}$  and  $\dim W = 3$ .

**Example 115.** Find bases for  $\text{col}(A)$  and  $\text{row}(A)$  with  $A = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$ .

**Solution.** Obviously,  $\text{col}(A) = \text{span}\left\{\begin{bmatrix} 0 \\ 1 \end{bmatrix}\right\}$  is 1-dimensional with basis  $\begin{bmatrix} 0 \\ 1 \end{bmatrix}$ .

Similarly,  $\text{row}(A) = \text{span}\left\{\begin{bmatrix} 1 & 0 \end{bmatrix}\right\}$  is 1-dimensional with basis  $\begin{bmatrix} 1 & 0 \end{bmatrix}$ .

During elimination, we usually do **row operations**. For instance, here,  $\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \xrightarrow[R_1 \leftrightarrow R_2]{\sim} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}$ .

- Note that the column spaces changed:  $\text{col}\left(\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}\right) \neq \text{col}\left(\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}\right)$
- But the row spaces did not change:  $\text{row}\left(\begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}\right) = \text{row}\left(\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}\right)$

Row operations preserve the row space but (generally) not the column space.

**Theorem 116.** Let  $A$  be a matrix and  $B$  an echelon form of  $A$ .

(a) The columns of  $A$  corresponding to the pivot columns of  $B$  form a basis for  $\text{col}(A)$ .

(b) The nonzero rows of  $B$  form a basis for  $\text{row}(A)$ .

(c) In particular, the dimensions of  $\text{col}(A)$  and  $\text{row}(A)$  are both equal to the number of pivots.

**Just for fun**

**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 pivot: <math>\implies</math> on the order of <math>N^2</math> operations</li> <li>• if there is <math>N</math> pivots: <math>\implies</math> on the order of <math>N \cdot N^2 = N^3</math> op's</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):  $N^{2.373}$

Is  $N^{2+(a \text{ tiny bit})}$  possible? **We have no idea!** (better is impossible; why?)

Good news for applications:

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