

### 1. Why Gauss elimination does not change the null space of matrix A?

let M be the product of elementary matrixes of A.

- suppose  $x \in N(A)$ , so  $Ax = 0$ . we have:

$$MAx = M(Ax) = M \cdot 0 = 0$$

so  $x \in N(MA)$  as well.

- suppose  $y \in N(MA)$ , so  $(MA)y = 0$ . we have:

$$Ay = M^{-1}MAy = M^{-1} \cdot 0 = 0$$

so  $y \in N(A)$  as well.

That means when you apply Gauss elimination, you do change the solution space.

### 2. Why every elementary is invertible?

[https://en.wikipedia.org/wiki/Elementary\\_matrix](https://en.wikipedia.org/wiki/Elementary_matrix)

<https://math.vanderbilt.edu/sapirmv/msapir/prleelementary.html>

### 3. If n vectors are independent, is their span $\mathbb{R}^n$ ?

Yes. Here is why:

**Proposition.** Let W be a subspace of a finite-dimensional vector space V. If  $\dim(W) = \dim(V)$ , then  $W=V$ .

If we are given linearly independent  $\{v_1, \dots, v_n\} \in \mathbb{R}^n$ , then  $\text{span}\{v_1, \dots, v_n\}$  is an n-dimensional subspace of  $\mathbb{R}^n$ . Since  $\dim(\mathbb{R}^n) = n$ , the proposition implies that  $\text{span}\{v_1, \dots, v_n\} = \mathbb{R}^n$

### 4. Proof: A invertible $\Leftrightarrow$ A's columns are independent & A is squared.

$A = [v_1, v_2, v_3, \dots, v_n] : v_i \text{ is column vector}$

- Suppose A is invertible  $\Rightarrow$  A is squared

$$Ax = 0 \Leftrightarrow A^{-1}Ax = A^{-1}0 \Leftrightarrow x = 0$$

$\Rightarrow$  q.e.d

- Suppose A's columns are independent & A is squared.

$\{v_1, v_2, \dots, v_n\}$  are independent  $\Rightarrow \{v_1, v_2, \dots, v_n\}$  is a basis of  $\mathbb{R}^n$  (see Q3). So:

$$I_n = A * \begin{bmatrix} b_{1,1} & \dots & b_{1,n} \\ b_{2,1} & \dots & b_{2,n} \\ \dots & & \\ b_{n,1} & \dots & b_{n,n} \end{bmatrix} = A * B \Rightarrow B = A^{-1}$$

### 5. Proof: A's columns are independent $\Rightarrow A^T A$ invertible.

- $A^T A$  is squared.
- A's columns are independent  $\Leftrightarrow N(A) = \{0\}$ . Let  $\vec{v} \in N(A^T A)$ . We have:

$$\begin{aligned}
& A^T A \cdot \vec{v} = \vec{0} \\
\iff & \vec{v}^T A^T A \vec{v} = \vec{v}^T \cdot \vec{0} = 0 \\
\iff & (A\vec{v})^T \cdot (A\vec{v}) = 0 \\
\iff & A\vec{v} = 0 \\
\iff & \begin{cases} \vec{v} = \vec{0} \\ \vec{v} \in N(A) \end{cases} \\
\iff & N(A) = N(A^T A) = \{0\}
\end{aligned}$$

=> q.e.d

## 6. Proof: Orthonormal set of vector are independent.

<https://www.quora.com/Is-an-orthonormal-set-of-vectors-a-linearly-independent-set/answer/Supreeth-Narasimhaswamy>

## 7. Why different eigenvalues produce independent eigenvectors?

Suppose we have  $k = 2$  eigenvalues and some combination of  $\mathbf{x}_1$  and  $\mathbf{x}_2$  produces 0:

$c_1 \mathbf{x}_1 + c_2 \mathbf{x}_2 = 0$  (1). We will prove that  $c_1 = c_2 = 0$ :

$A c_1 \mathbf{x}_1 + A c_2 \mathbf{x}_2 = 0 \iff c_1 \lambda_1 \mathbf{x}_1 + c_2 \lambda_2 \mathbf{x}_2 = 0$  (2)

Subtract (1) $\lambda_2$  to (2):

$\mathbf{x}_1 c_1 (\lambda_1 - \lambda_2) = 0 \iff c_1 = 0$  because  $\mathbf{x}_1 \neq 0$  and eigenvalues are different.

Similarly, we can show that  $c_2 = 0$ . This same argument extends to any number of eigenvalue.

## 8. Why A is singular $\iff$ A has an eigenvalue equal 0?

Because:  $\det(A) = \lambda_1 \cdot \dots \cdot \lambda_n$

**Eigenvectors of a symmetric matrix A corresponding to different eigenvalues are orthogonal.**

<http://www.maths.manchester.ac.uk/~peter/MATH10212/notes10.pdf>