

## The Rank–Nullity Theorem and the Four Fundamental Subspaces

### Finding Bases for the Four Fundamental Subspaces via Row Echelon Reduction

We assume that you know how to perform Gaussian elimination to reduce a matrix to Row Echelon form. Let  $A$  be an  $m \times n$  matrix. Perform Gaussian elimination to reduce  $A$  to its (reduced) row echelon form  $R$ . The pivot columns and rows of  $R$  provide the necessary information to construct bases for all four fundamental subspaces.

The first step is to use elementary row operations to obtain a matrix  $R$  in (reduced) row echelon form that is row-equivalent to  $A$ . Let the pivot columns be indexed by  $j_1, j_2, \dots, j_r$ , where  $r = \text{rank}(A)$ .

#### Basis for the Row Space $\text{Row}(A)$

The nonzero rows of  $R$  form a basis for  $\text{Row}(A)$ . This is because elementary row operations do not change the row space.

#### Basis for the Column space $\text{Col}(A)$

Let  $A$  be an  $m \times n$  real matrix. We aim to explain why the columns of the **original matrix**  $A$  that correspond to the **pivot columns** in a row echelon form of  $A$  constitute a basis for the column space  $\text{Col}(A)$ .

Begin by writing the matrix  $A$  in terms of its columns:

$$A = [\mathbf{a}_1 \quad \mathbf{a}_2 \quad \cdots \quad \mathbf{a}_n],$$

where each  $\mathbf{a}_j \in \mathbb{R}^m$ . The **column space** of  $A$  is the subspace of  $\mathbb{R}^m$  defined by

$$\text{Col}(A) = \text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\}.$$

A **basis** for  $\text{Col}(A)$  is a subset of these columns that is

- (i) linearly independent, and

(ii) spans  $\text{Col}(A)$ .

The next step is to row reduce  $A$  to row echelon form. Perform Gaussian elimination on  $A$  to obtain a row echelon form  $R$ . Since elementary row operations are invertible, there exists an invertible  $m \times m$  matrix  $E$  such that

$$R = EA.$$

Let the pivot columns of  $R$  be in positions  $j_1, j_2, \dots, j_r$ , where  $r = \text{rank}(A)$ . These are the columns that contain the leading (first nonzero) entries of each nonzero row in  $R$ .

### 3. Properties of pivot columns in row echelon form

In the row echelon matrix  $R$ , the following hold:

- The pivot columns  $\mathbf{r}_{j_1}, \dots, \mathbf{r}_{j_r}$  are **linearly independent**.
- Every non-pivot column of  $R$  is a **linear combination** of the pivot columns to its left.

These facts follow from the staircase structure of  $R$ : each pivot appears in a new row below the previous one, and all entries below a pivot are zero. This allows back-substitution to verify independence and express non-pivot columns as combinations of earlier pivot columns.

Thus,  $\{\mathbf{r}_{j_1}, \dots, \mathbf{r}_{j_r}\}$  is a basis for  $\text{Col}(R)$ .

**However**, note that in general

$$\text{Col}(A) \neq \text{Col}(R),$$

so we **cannot** use the columns of  $R$  as a basis for  $\text{Col}(A)$ .

Although the column spaces differ, row operations **preserve all linear dependence relations among the columns**. To see this, observe:

$$A\mathbf{x} = \mathbf{0} \iff EA\mathbf{x} = E\mathbf{0} \iff R\mathbf{x} = \mathbf{0}.$$

Hence,

$$\text{Nul}(A) = \text{Nul}(R).$$

This equality means that a set of columns of  $A$  is linearly dependent **if and only if** the corresponding set of columns of  $R$  is linearly dependent. In

other words, the *pattern* of linear dependence among the columns is identical in  $A$  and  $R$ .

Because the dependence relations are the same:

- The columns  $\mathbf{a}_{j_1}, \dots, \mathbf{a}_{j_r}$  of  $A$  (in the same positions as the pivot columns of  $R$ ) are **linearly independent**.

*Reason:* If they were linearly dependent, the same dependence would appear among  $\mathbf{r}_{j_1}, \dots, \mathbf{r}_{j_r}$ , contradicting their independence in  $R$ .

- Every non-pivot column  $\mathbf{a}_k$  of  $A$  (where  $k \notin \{j_1, \dots, j_r\}$ ) is a **linear combination** of  $\mathbf{a}_{j_1}, \dots, \mathbf{a}_{j_r}$ .

*Reason:* In  $R$ , we have  $\mathbf{r}_k = c_1\mathbf{r}_{j_1} + \dots + c_r\mathbf{r}_{j_r}$  for some scalars  $c_i$ . Then the vector  $\mathbf{x}$  with  $x_k = 1$ ,  $x_{j_i} = -c_i$ , and other entries zero satisfies  $R\mathbf{x} = \mathbf{0}$ . Since  $\text{Nul}(A) = \text{Nul}(R)$ , we also have  $A\mathbf{x} = \mathbf{0}$ , which implies  $\mathbf{a}_k = c_1\mathbf{a}_{j_1} + \dots + c_r\mathbf{a}_{j_r}$ .

Therefore, the set

$$\{\mathbf{a}_{j_1}, \mathbf{a}_{j_2}, \dots, \mathbf{a}_{j_r}\}$$

is linearly independent and spans  $\text{Col}(A)$ .

## Conclusion

This set is a **basis for the column space of  $A$** . Hence:

**Theorem 1.** *Let  $A$  be an  $m \times n$  matrix, and let  $R$  be any row echelon form of  $A$ . If the pivot columns of  $R$  occur in positions  $j_1, \dots, j_r$ , then the corresponding columns of the original matrix  $A$ ,*

$$\{\mathbf{a}_{j_1}, \dots, \mathbf{a}_{j_r}\},$$

*form a basis for  $\text{Col}(A)$ .*

**Remark 1.** *It is crucial to use the columns of the **original matrix**  $A$ , not those of  $R$ . While  $R$  reveals which columns to select, the actual basis vectors must come from  $A$  because  $\text{Col}(A) \neq \text{Col}(R)$  in general.*

## Example

Given a matrix  $A$ , find a basis for its column space  $\text{Col}(A)$  — that is, a linearly independent set of columns of  $A$  that spans  $\text{Col}(A)$ .

Let

$$A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 2 & 4 & 6 & 2 \\ 1 & 0 & 1 & 2 \\ 0 & 2 & 4 & 1 \end{bmatrix}.$$

This is a  $4 \times 4$  matrix. Its column space is a subspace of  $\mathbb{R}^4$ .

### Step-by-Step Procedure

**Step 1: Row reduce  $A$  to row echelon form (REF)**

Perform Gaussian elimination:

$$\begin{aligned} A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 2 & 4 & 6 & 2 \\ 1 & 0 & 1 & 2 \\ 0 & 2 & 4 & 1 \end{bmatrix} &\xrightarrow{R_2 \leftarrow R_2 - 2R_1, R_3 \leftarrow R_3 - R_1} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & -2 & -2 & 1 \\ 0 & 2 & 4 & 1 \end{bmatrix} \\ &\xrightarrow{\text{swap } R_2 \leftrightarrow R_3} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -2 & -2 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 2 & 4 & 1 \end{bmatrix} \xrightarrow{R_4 \leftarrow R_4 + R_2} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -2 & -2 & 1 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 2 \end{bmatrix} \\ &\xrightarrow{\text{swap } R_3 \leftrightarrow R_4} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -2 & -2 & 1 \\ 0 & 0 & 2 & 2 \\ 0 & 0 & 0 & 0 \end{bmatrix}. \end{aligned}$$

This is a row echelon form (REF) of  $A$ .

**Step 2: Identify the pivot columns**

The leading (first nonzero) entry in each nonzero row occurs in:

- Row 1: column 1,
- Row 2: column 2,
- Row 3: column 3. Thus, the **pivot columns** are columns 1, 2, and 3. Hence,  $\text{rank}(A) = 3$ .

**Step 3: Select the corresponding columns from the original matrix  $A$**

Take columns 1, 2, and 3 of the **original** matrix  $A$ :

$$\mathbf{a}_1 = \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \quad \mathbf{a}_2 = \begin{bmatrix} 2 \\ 4 \\ 0 \\ 2 \end{bmatrix}, \quad \mathbf{a}_3 = \begin{bmatrix} 3 \\ 6 \\ 1 \\ 4 \end{bmatrix}.$$

**Step 4: Justification**

- These columns are **linearly independent** because the corresponding columns in the REF are linearly independent, and row operations preserve linear dependence relations among columns.
- They **span**  $\text{Col}(A)$  because every non-pivot column (here, column 4) is a linear combination of the pivot columns, and the same linear relation holds in  $A$  as in its REF.

**Demonstration: Identical Linear Dependence Relations in  $A$  and  $R$**

Let us verify explicitly that linear dependence relations among columns are the same in  $A$  and  $R$ .

Denote the columns of  $R$  by  $\mathbf{r}_1, \mathbf{r}_2, \mathbf{r}_3, \mathbf{r}_4$ . From the REF:

$$R = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -2 & -2 & 1 \\ 0 & 0 & 2 & 2 \\ 0 & 0 & 0 & 0 \end{bmatrix},$$

so

$$\mathbf{r}_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{r}_2 = \begin{bmatrix} 2 \\ -2 \\ 0 \\ 0 \end{bmatrix}, \quad \mathbf{r}_3 = \begin{bmatrix} 3 \\ -2 \\ 2 \\ 0 \end{bmatrix}, \quad \mathbf{r}_4 = \begin{bmatrix} 1 \\ 1 \\ 2 \\ 0 \end{bmatrix}.$$

Observe that column 4 of  $R$  can be expressed as a linear combination of columns 1–3. Solving

$$c_1\mathbf{r}_1 + c_2\mathbf{r}_2 + c_3\mathbf{r}_3 = \mathbf{r}_4,$$

we work from the bottom up (using the echelon structure):

- Row 3:  $0c_1 + 0c_2 + 2c_3 = 2 \Rightarrow c_3 = 1$ , - Row 2:  $0c_1 - 2c_2 - 2c_3 = 1 \Rightarrow -2c_2 - 2(1) = 1 \Rightarrow c_2 = -\frac{3}{2}$ , - Row 1:  $c_1 + 2c_2 + 3c_3 = 1 \Rightarrow c_1 + 2(-\frac{3}{2}) + 3(1) = 1 \Rightarrow c_1 - 3 + 3 = 1 \Rightarrow c_1 = 1$ .

Thus,

$$\mathbf{r}_4 = 1 \cdot \mathbf{r}_1 - \frac{3}{2} \cdot \mathbf{r}_2 + 1 \cdot \mathbf{r}_3.$$

Now consider the same linear combination of the \*\*original columns\*\* of  $A$ :

$$1 \cdot \mathbf{a}_1 - \frac{3}{2} \cdot \mathbf{a}_2 + 1 \cdot \mathbf{a}_3 = \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix} - \frac{3}{2} \begin{bmatrix} 2 \\ 4 \\ 0 \\ 2 \end{bmatrix} + \begin{bmatrix} 3 \\ 6 \\ 1 \\ 4 \end{bmatrix} = \begin{bmatrix} 1 - 3 + 3 \\ 2 - 6 + 6 \\ 1 - 0 + 1 \\ 0 - 3 + 4 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \\ 2 \\ 1 \end{bmatrix}.$$

But this is exactly the fourth column of  $A$ :

$$\mathbf{a}_4 = \begin{bmatrix} 1 \\ 2 \\ 2 \\ 1 \end{bmatrix}.$$

Hence,

$$\mathbf{a}_4 = 1 \cdot \mathbf{a}_1 - \frac{3}{2} \cdot \mathbf{a}_2 + 1 \cdot \mathbf{a}_3.$$

The \*\*same coefficients\*\* that express  $\mathbf{r}_4$  as a combination of  $\mathbf{r}_1, \mathbf{r}_2, \mathbf{r}_3$  also express  $\mathbf{a}_4$  as a combination of  $\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3$ .

This confirms that \*\*linear dependence relations among columns are identical in  $A$  and  $R$ \*\*, which is why the pivot columns of  $A$  are linearly independent and span  $\text{Col}(A)$ .

## Final Answer

A basis for  $\text{Col}(A)$  is

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ 0 \\ 2 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \\ 1 \\ 4 \end{bmatrix} \right\}.$$

This set contains 3 vectors (matching  $\text{rank}(A) = 3$ ), is linearly independent, and spans  $\text{Col}(A)$ .

**Remark 2. Important:** Always use the columns of the *original matrix*  $A$ , not the columns of the row echelon form. The column space of the REF is generally different from that of  $A$ .

## Right Null Space

### Goal

Given an  $m \times n$  matrix  $A$ , find a **basis** for its null space

$$\text{Nul}(A) = \{\mathbf{x} \in \mathbb{R}^n : A\mathbf{x} = \mathbf{0}\},$$

i.e., a linearly independent set of vectors in  $\mathbb{R}^n$  whose span is exactly the set of all solutions to  $A\mathbf{x} = \mathbf{0}$ .

### Core Idea

The null space consists of all vectors  $\mathbf{x}$  such that the linear combination of the columns of  $A$  with coefficients  $x_1, \dots, x_n$  equals the zero vector:

$$x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \cdots + x_n\mathbf{a}_n = \mathbf{0}.$$

Row reduction does not change the solution set of this homogeneous system. Thus, we may solve the simpler system  $R\mathbf{x} = \mathbf{0}$ , where  $R$  is the (reduced) row echelon form of  $A$ .

## Step-by-Step Procedure

### Step 1: Reduce $A$ to Reduced Row Echelon Form (RREF)

Use Gaussian elimination with back-substitution to obtain the **reduced row echelon form**  $R$  of  $A$ .

Since elementary row operations are invertible and preserve the solution set,

$$A\mathbf{x} = \mathbf{0} \quad \Longleftrightarrow \quad R\mathbf{x} = \mathbf{0}.$$

### Step 2: Identify Pivot and Free Columns

In  $R$ :

- A **pivot column** contains a leading 1 (the first nonzero entry in its row).
- A **free column** has no pivot.

Let:

- $r = \text{rank}(A) = \text{number of pivot columns}$ ,
- $n - r = \text{number of free columns}$ .

The variables corresponding to:

- **Pivot columns** are called **basic variables**,
- **Free columns** are called **free variables**.

The free variables can be assigned arbitrary values; the basic variables are then uniquely determined.

### Step 3: Solve $R\mathbf{x} = \mathbf{0}$ in Terms of Free Variables

Write the system  $R\mathbf{x} = \mathbf{0}$  as equations. Because  $R$  is in RREF, each nonzero row gives an equation of the form

$$x_{\text{pivot}} + \sum_{\text{free } j > \text{pivot}} r_{ij}x_j = 0,$$

so

$$x_{\text{pivot}} = - \sum_{\text{free } j > \text{pivot}} r_{ij}x_j.$$

Thus, every basic variable is expressed as a linear combination of the free variables.

### Step 4: Construct Special Solutions (Basis Vectors)

For each free variable  $x_{f_k}$  (where  $k = 1, 2, \dots, n - r$ ), construct a vector  $\mathbf{v}_k \in \mathbb{R}^n$  as follows:

- Set  $x_{f_k} = 1$ .
- Set all other free variables to 0.
- Use the equations from Step 3 to compute the values of all basic variables.
- Assemble the full vector  $\mathbf{v}_k = (x_1, x_2, \dots, x_n)^\top$ .

Each  $\mathbf{v}_k$  is called a **special solution** corresponding to the free variable  $x_{f_k}$ .



**Step 5: The Set  $\{\mathbf{v}_1, \dots, \mathbf{v}_{n-r}\}$  is a Basis**

- **Linear independence:** In the subvector corresponding to free variables,  $\mathbf{v}_k$  has a 1 in position  $f_k$  and 0 elsewhere. Hence, no nontrivial linear combination can be zero.
- **Spanning:** Any solution  $\mathbf{x}$  to  $A\mathbf{x} = \mathbf{0}$  is determined by its free variable values  $c_1, \dots, c_{n-r}$ , and then

$$\mathbf{x} = c_1\mathbf{v}_1 + \dots + c_{n-r}\mathbf{v}_{n-r}.$$

Therefore, this set is a basis for  $\text{Nul}(A)$ .

**Worked Example**

Let

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 2 & 4 & 6 & 8 \\ 1 & 1 & 2 & 3 \end{bmatrix}.$$

**Step 1: Reduce to RREF**

$$A \xrightarrow{\text{row reduce}} R = \begin{bmatrix} 1 & 0 & 1 & 2 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

**Step 2: Identify pivot and free columns**

- Pivot columns: 1 and 2  $\Rightarrow$  basic variables:  $x_1, x_2$ ,
- Free columns: 3 and 4  $\Rightarrow$  free variables:  $x_3, x_4$ ,
- Rank  $r = 2$ , so  $\dim(\text{Nul}(A)) = 4 - 2 = 2$ .

**Step 3: Solve  $R\mathbf{x} = \mathbf{0}$**

From  $R$ :

$$\begin{cases} x_1 + x_3 + 2x_4 = 0 \\ x_2 + x_3 + x_4 = 0 \end{cases} \Rightarrow \begin{cases} x_1 = -x_3 - 2x_4 \\ x_2 = -x_3 - x_4 \end{cases}$$

#### Step 4: Construct special solutions

- For  $x_3 = 1, x_4 = 0$ :

$$x_1 = -1, \quad x_2 = -1 \quad \Rightarrow \quad \mathbf{v}_1 = \begin{bmatrix} -1 \\ -1 \\ 1 \\ 0 \end{bmatrix}.$$

- For  $x_3 = 0, x_4 = 1$ :

$$x_1 = -2, \quad x_2 = -1 \quad \Rightarrow \quad \mathbf{v}_2 = \begin{bmatrix} -2 \\ -1 \\ 0 \\ 1 \end{bmatrix}.$$

### Left Null Space

Let  $A$  be an  $m \times n$  real matrix. The **left null space** of  $A$  is defined as

$$\text{Nul}(A^\top) = \{\mathbf{y} \in \mathbb{R}^m : A^\top \mathbf{y} = \mathbf{0}\} = \{\mathbf{y} \in \mathbb{R}^m : \mathbf{y}^\top A = \mathbf{0}^\top\}.$$

Its dimension is  $m - r$ , where  $r = \text{rank}(A)$ . We now describe a systematic method to compute a basis for this subspace using Gaussian elimination.

### Method: Using the Elimination Matrix

The key idea is to record the row operations used to reduce  $A$  to reduced row echelon form (RREF). This is done by augmenting  $A$  with the identity matrix and performing simultaneous row reduction.

#### Step 1: Form the augmented matrix

Construct the  $m \times (n + m)$  matrix

$$[A \mid I_m],$$

where  $I_m$  is the  $m \times m$  identity matrix.

**Step 2: Row-reduce to RREF**

Apply Gaussian elimination (with back-substitution to achieve reduced form) to the left block, performing the same elementary row operations on the entire augmented matrix. The result is

$$[R \mid E],$$

where:

- $R$  is the reduced row echelon form of  $A$ ,
- $E$  is an  $m \times m$  invertible matrix satisfying  $EA = R$ .

**Step 3: Determine the rank**

Let  $r = \text{rank}(A)$  be the number of nonzero rows in  $R$ . By the definition of RREF,  $R$  has the block structure

$$R = \begin{bmatrix} R_1 \\ \mathbf{0} \end{bmatrix},$$

where  $R_1$  is  $r \times n$  and  $\mathbf{0}$  is  $(m - r) \times n$ .

**Step 4: Partition  $E$** 

Write  $E$  in conformal block form:

$$E = \begin{bmatrix} E_1 \\ E_2 \end{bmatrix},$$

where  $E_1$  is  $r \times m$  and  $E_2$  is  $(m - r) \times m$  (the last  $m - r$  rows of  $E$ ).

Since  $EA = R$ , we have

$$\begin{bmatrix} E_1 A \\ E_2 A \end{bmatrix} = \begin{bmatrix} R_1 \\ \mathbf{0} \end{bmatrix} \implies E_2 A = \mathbf{0}.$$

**Step 5: Interpret  $E_2 A = \mathbf{0}$** 

The equation  $E_2 A = \mathbf{0}$  means that each row  $\mathbf{y}^\top$  of  $E_2$  satisfies

$$\mathbf{y}^\top A = \mathbf{0}^\top \iff A^\top \mathbf{y} = \mathbf{0}.$$

Thus, every row of  $E_2$  (as a column vector) lies in  $\text{Nul}(A^\top)$ .

**Step 6: Conclude a basis**

- $E_2$  has  $m - r$  rows.
- Since  $E$  is invertible, its rows are linearly independent; hence the rows of  $E_2$  are linearly independent.
- $\dim(\text{Nul}(A^\top)) = m - r$  (by the Rank–Nullity Theorem applied to  $A^\top$ ).

Therefore, the rows of  $E_2$ , when written as column vectors, form a basis for  $\text{Nul}(A^\top)$ .

**Worked Example**

Let

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & 6 \\ 1 & 1 & 2 \end{bmatrix}.$$

**Step 1:** Form  $[A \mid I_3]$ :

$$[A \mid I_3] = \left[ \begin{array}{ccc|ccc} 1 & 2 & 3 & 1 & 0 & 0 \\ 2 & 4 & 6 & 0 & 1 & 0 \\ 1 & 1 & 2 & 0 & 0 & 1 \end{array} \right].$$

**Step 2:** Row-reduce to RREF:

$$[R \mid E] = \left[ \begin{array}{ccc|ccc} 1 & 0 & 1 & -1 & 0 & 2 \\ 0 & 1 & 1 & 1 & 0 & -1 \\ 0 & 0 & 0 & -2 & 1 & 0 \end{array} \right].$$

**Step 3:** Rank  $r = 2$ , so  $m - r = 1$ .

**Step 4:** The last row of  $E$  is  $[-2 \ 1 \ 0]$ .

**Step 5:** Verify:

$$[-2 \ 1 \ 0]A = [0 \ 0 \ 0].$$

**Step 6:** A basis for  $\text{Nul}(A^\top)$  is

$$\left\{ \begin{bmatrix} -2 \\ 1 \\ 0 \end{bmatrix} \right\}.$$

## Summary of the Procedure

To find a basis for  $\text{Nul}(A^\top)$ :

1. Form the augmented matrix  $[A \mid I_m]$ .
2. Row-reduce the left block to RREF, applying the same operations to the right block to obtain  $[R \mid E]$ .
3. Let  $r = \text{rank}(A)$  (number of nonzero rows in  $R$ ).
4. Extract the last  $m - r$  rows of  $E$ .
5. Transpose each of these rows to column vectors; the resulting set is a basis for  $\text{Nul}(A^\top)$ .

**Remark 3.** *The left null space consists of all linear combinations of the rows of  $A$  that yield the zero vector. The last  $m - r$  rows of  $E$  encode precisely those combinations, which is why they form a basis.*

**Remark 4.** *Do **not** use the zero rows of  $R$ —they are trivial. The useful information is stored in the corresponding rows of  $E$ .*

### Step 5: Basis for the Left Null Space $\text{Nul}(A^\top)$

Let  $A$  be an  $m \times n$  real matrix. The **left null space** of  $A$  is defined as

$$\text{Nul}(A^\top) = \{\mathbf{y} \in \mathbb{R}^m : A^\top \mathbf{y} = \mathbf{0}\} = \{\mathbf{y} \in \mathbb{R}^m : \mathbf{y}^\top A = \mathbf{0}^\top\}.$$

Its dimension is  $m - r$ , where  $r = \text{rank}(A)$ . We now describe a systematic method to compute a basis for this subspace using Gaussian elimination.

**Remark 5.** *The left null space consists of all linear combinations of the rows of  $A$  that yield the zero vector. The last  $m - r$  rows of  $E$  encode precisely those combinations, which is why they form a basis.*

**Remark 6.** *Do **not** use the zero rows of  $R$ —they are trivial. The useful information is stored in the corresponding rows of  $E$ .*