



Matrix Rank

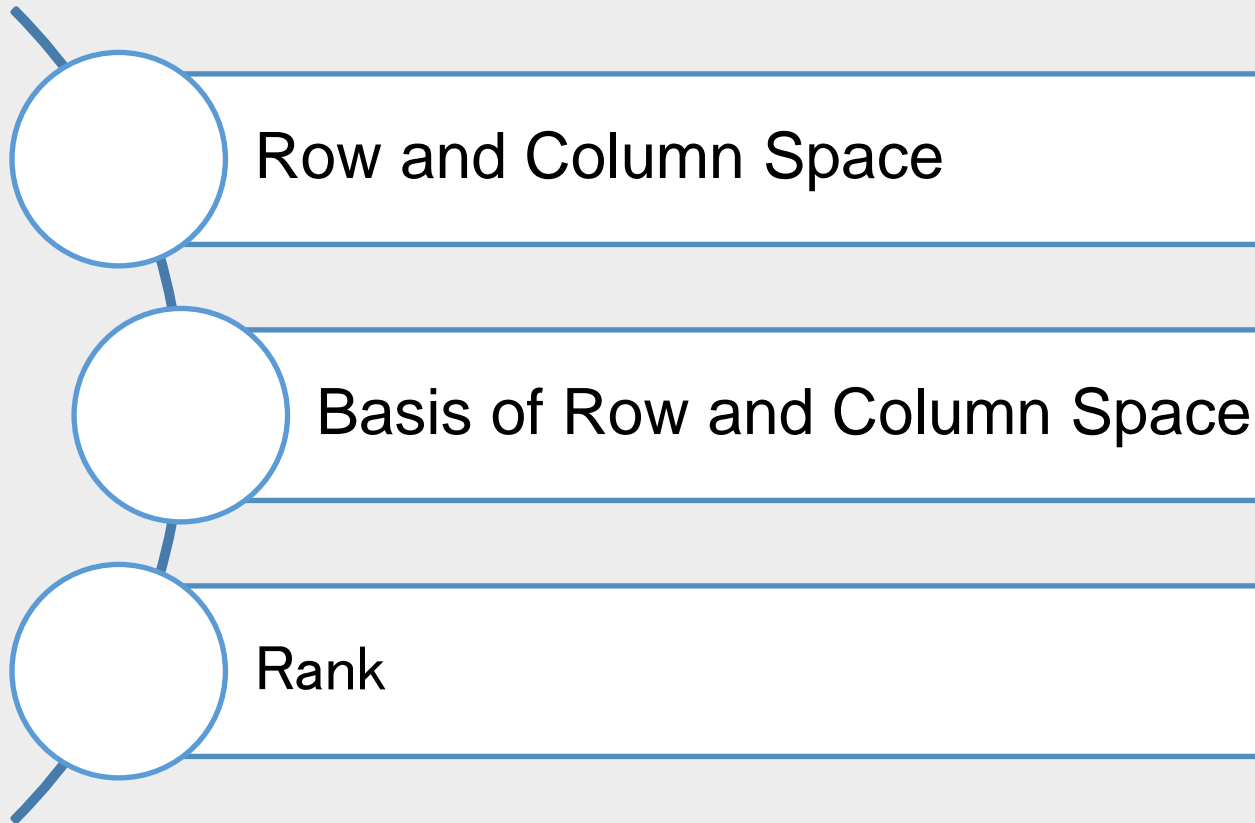
Linear Algebra

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Row and Column Space



Definition

Let A be a $m \times n$ matrix. Then the **column space** of A is $\mathcal{C}(A)$ is

$$\mathcal{C}(A) := \{Ax : x \in \mathbb{R}^n\}$$

and the **row space** of A is

$$\mathcal{R}(A) := \{y^T A : y \in \mathbb{R}^m\}.$$



The *row space* of a matrix is the collection of all linear combinations of its rows.

Equivalently, the row space is the span of rows.

The elements of a row space are *row vectors*.

If a matrix has m columns, its row space is a subspace of (the row version of) \mathbb{R}^m

Elementary row operations **do not** alter the row space.

Thus a matrix and its echelon form have the same row space.

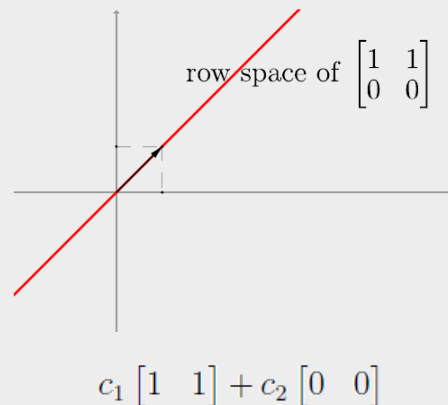
The pivot rows of an echelon form are **linearly independent**.

$$\begin{bmatrix} 1 & * & * & * & * \\ & 1 & * & * & * \\ & & 1 & * & * \end{bmatrix}$$

The pivot rows of an echelon form span the row space of the original matrix.

The dimension of the row space is given by the number of pivot rows.

This dimension does not exceed the total row count.





The *column space* of a matrix is the collection of all linear combinations of its columns.

It is the span of columns, the range of the linear transformation carried out by the matrix.

If a matrix has n rows, its column space is a subspace of \mathbb{R}^n .

Elementary row operations **affect** the column space.

So, generally, a matrix and its echelon form have different column spaces.

However, since the row operations preserve the linear relations between columns,
the columns of an echelon form and the original columns obey the *same* relations.

Example

The pivot columns of a reduced row-echelon form are **linearly independent**.

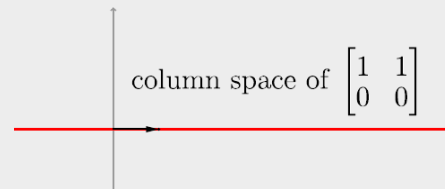
$$\begin{bmatrix} 1 & * & & * \\ & & 1 & * \\ & & & 1 & * \end{bmatrix}$$

The pivot columns of a reduced row-echelon form **span** its column space.

So the pivot columns of a matrix are linearly independent and span its column space.

The dimension of the column space is given by the number of pivot columns.

This dimension does not exceed the total column count.



$$c_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$



$$B = \begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & 5 & 8 & 11 & 14 \\ 1 & 3 & 5 & 8 & 11 \\ 4 & 10 & 16 & 23 & 30 \end{pmatrix} \quad B_{\text{rref}} = \begin{pmatrix} 1 & 0 & -1 & 0 & 1 \\ 0 & 1 & 2 & 0 & -2 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\begin{aligned} x_1 &= x_3 - x_5, \\ x_2 &= -2x_3 + 2x_5, \\ x_4 &= -2x_5. \end{aligned} \quad \mathbf{x} = x_3 \begin{pmatrix} 1 \\ -2 \\ 1 \\ 0 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} -1 \\ 2 \\ 0 \\ -2 \\ 1 \end{pmatrix}.$$

The column space of B is 3-dimensional, and that a basis is given by $\left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \\ 4 \end{bmatrix}, \begin{bmatrix} 2 \\ 5 \\ 3 \\ 10 \end{bmatrix}, \begin{bmatrix} 4 \\ 11 \\ 8 \\ 23 \end{bmatrix} \right\}$

Note that we do not use the columns of B_{rref} ! We use the columns of B .

Basis of Row and Column Space



Theorem

If two matrices A and B are row-equivalent, then their row spaces are the same. If B is in echelon form, the **non-zero rows (pivot rows)** of B form a basis for the row space of A as well as for that of B .



Definition

□ If a matrix in echelon form satisfies the following additional conditions, then it is in **reduced echelon form** (or **reduced row echelon form**):

1. The leading entry in each non-zero row is 1.
2. Each leading 1 is the only non-zero entry in its columns.
3. The leading 1 in the second row or beyond is to the right of the leading 1 in the row just above.
4. Any row containing only 0's is at the bottom.

$$\begin{array}{c} e_1 \quad e_2 \quad e_3 \\ \left[\begin{array}{ccc|c} 1 & 0 & 0 & 29 \\ 0 & 1 & 0 & 16 \\ 0 & 0 & 1 & 3 \end{array} \right] \end{array}$$

Reduced Echelon form



Theorem

The pivot columns of a matrix A form a basis for $Col(A)$

- Lemma1: The pivot columns of A are linearly independent
- Lemma 2. The pivot columns of A span the column space of A

$$\begin{bmatrix} 1 & b_{12} & 0 & b_{14} & 0 & b_{16} \\ 0 & 0 & 1 & b_{24} & 0 & b_{26} \\ 0 & 0 & 0 & 0 & 1 & b_{36} \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$



span the column space.

We use the reduced echelon matrix of \mathbf{A} in the proof. We designate it by \mathbf{B} . If \mathbf{A} and \mathbf{B} have r pivot columns, then the pivot columns of \mathbf{B} are the standard vectors $\mathbf{e}^1, \dots, \mathbf{e}^r$ in \mathbb{R}^m . For example, consider the reduced echelon matrix

$$\begin{bmatrix} 1 & b_{12} & 0 & b_{14} & 0 & b_{16} \\ 0 & 0 & 1 & b_{24} & 0 & b_{26} \\ 0 & 0 & 0 & 0 & 1 & b_{36} \\ 0 & 0 & 0 & 0 & 0 & 0 \end{bmatrix},$$

which has pivot columns \mathbf{e}^1 , \mathbf{e}^2 , and \mathbf{e}^3 in its first, third, and fourth columns. These are vectors in \mathbb{R}^4 .

The solutions of the homogeneous equations $\mathbf{Ax} = \mathbf{0}$ and $\mathbf{Bx} = \mathbf{0}$ are the same. Therefore, the relations among the columns of \mathbf{A} and \mathbf{B} are the same. (Any non-zero solution of these homogeneous equations gives a relation among these columns.)

Lemma 1. (a) *The pivot columns of \mathbf{B} are linearly independent.*

(b) *The pivot columns of \mathbf{A} are linearly independent.*

Proof. (a) The pivot columns of \mathbf{B} are the standard vectors $\mathbf{e}^1, \dots, \mathbf{e}^r$ in \mathbb{R}^m that are linearly independent. (In the example of the matrix given above, \mathbf{e}^1 , \mathbf{e}^2 , and \mathbf{e}^3 are linearly independent in \mathbb{R}^4 .)

(b) The relations among the columns of \mathbf{A} and \mathbf{B} are the same (as noted above), so the pivot columns of \mathbf{A} are also linearly independent. \square

Lemma 2. *The pivot columns of \mathbf{A} span the column space of \mathbf{A}*

Proof. Let \mathbf{b}^k be a non-pivot column of the reduced echelon matrix \mathbf{B} . Assume that there are j pivot columns to the left of \mathbf{b}^k in \mathbf{B} . These pivot columns must be $\mathbf{e}^1, \dots, \mathbf{e}^j$. The non-pivot column \mathbf{b}^k can only have nonzero entries in the first j components (or it would be a pivot column) and so is a linear combination of $\mathbf{e}^1, \dots, \mathbf{e}^j$.

Since \mathbf{A} has the same relations among its columns as \mathbf{B} , its non-pivot column \mathbf{a}^k is a linear combination of the j pivot columns to the left of it. By Theorem 5(a), the span of the pivot columns is the same as the span of all the columns. This shows that the pivot columns of \mathbf{A} span the column space of \mathbf{A} . \square

By Lemma 1(b), the pivot columns are linearly independent. By Lemma 2, the pivot columns span the column space of \mathbf{A} . Together, these two facts show that the pivot columns form a basis for the column space of \mathbf{A} . \square



Example

Find:

- ☐ Row Basis
- ☐ Column Basis
- ☐ $\dim(\text{Row}(A))$
- ☐ $\dim(\text{Col}(A))$
- ☐ $\dim(\text{Null}(A))$

$$A = \begin{bmatrix} -2 & -5 & 8 & 0 & -17 \\ 1 & 3 & -5 & 1 & 5 \\ 3 & 11 & -19 & 7 & 1 \\ 1 & 7 & -13 & 5 & -3 \end{bmatrix}$$

$$A \sim B = \begin{bmatrix} 1 & 3 & -5 & 1 & 5 \\ 0 & 1 & -2 & 2 & -7 \\ 0 & 0 & 0 & -4 & 20 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$A \sim B \sim C = \begin{bmatrix} 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & -2 & 0 & 3 \\ 0 & 0 & 0 & 1 & -5 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Number of non-zero rows = pivot columns

Rank



Definition

- We call $\dim(R(A))$ the **row rank** of A and $\dim(C(A))$ the **column rank** of A .
- We refer to a basis of $C(A)$ consisting of columns of A as a **column basis**.
A **row basis** is defined similarly.



Definition

- ❑ The number of linearly independent rows or columns in the matrix
- ❑ **Dimension of the row (column) space**
- ❑ Number of nonzero rows of the matrix in row echelon form (Ref)

Note

The dimension of the Column Space of A and $\text{rref}(A)$ is the same.



Example 1

If Columns of matrix A are linearly independent:

$$\text{nullity}(A) = ?$$

$$\text{colrank}(A) = ?$$



Example 2

$$A = \begin{bmatrix} 0 & 1 & 1 & 2 \\ 1 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, Ax = \begin{bmatrix} x_2 + x_3 + 2x_4 \\ x_1 + 2x_3 + x_4 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow x = \begin{bmatrix} -2x_3 - x_4 \\ -x_3 - 2x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} -2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -1 \\ -2 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{nullity}(A) = 2, \text{colRank}(A) = 2$$



Note

The dimension of $Null(A)$ is the number of free variables in the equation $Ax = 0$, and the dimension of $Col(A)$ is the number of pivot columns in A .

Example

Back to [Example!](#)

Find the dimension of the Null Space and the Column Space of

$$A = \begin{bmatrix} -3 & 6 & -1 & 1 & -7 \\ 1 & -2 & 2 & 3 & -1 \\ 2 & -4 & 5 & 8 & -4 \end{bmatrix}$$

(row reduce the Augmented Matrix $[A \ 0]$ to echelon form)



Important Note!!! (WHY?)

Column Space (A) = Range(A)

$\text{Dim}(\text{range}(A)) = \text{ColumnRank}(A)$

Theorem

$$\square \text{Nullity}(A) + \text{ColRank}(A) = n$$

$$\square \text{Dim}(\text{Null}(A)) + \text{Dim}(\text{Range}(A)) = n$$

$$\{\text{number of pivot columns}\} + \{\text{number of non – pivot columns}\} = \{\text{number of columns}\}$$



Theorem

□ Let A denote an $m \times n$ matrix of rank r . Then:

the $n-r$ basic solutions to the system $Ax=0$ provided by the gaussian algorithm are a basis of $\text{null}(A)$, so $\dim[\text{null}(A)] = n-r$.

Example

$$A = \begin{bmatrix} 0 & 1 & 1 & 2 \\ 1 & 0 & 2 & 1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix}, Ax = \begin{bmatrix} x_2 + x_3 + 2x_4 \\ x_1 + 2x_3 + x_4 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \Rightarrow x = \begin{bmatrix} -2x_3 - x_4 \\ -x_3 - 2x_4 \\ x_3 \\ x_4 \end{bmatrix} = x_3 \begin{bmatrix} -2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -1 \\ -2 \\ 0 \\ 1 \end{bmatrix}$$

$$\text{nullity}(A) = 2, \text{colRank}(A) = 2 \quad \text{Nullspace}(A) = \left\{ \begin{bmatrix} -2 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ -2 \\ 0 \\ 1 \end{bmatrix} \right\}$$



Theorem (RMRT)

(Rank of a matrix is equal to the rank of its transpose)

Suppose A is an $m \times n$ matrix.

Then $\text{colrank}(A) = \text{colrank}(A^T)$

Theorem

□ $\text{ColRank}(A) = \text{RowRank}(A)$

□ In general, It's called rank of matrix ($\text{rank}(A)$)

Proof?



1. As a set of vector–vector products

$$C = AB = \begin{bmatrix} - & a_1^T & - \\ - & a_2^T & - \\ & \vdots & \\ - & a_m^T & - \end{bmatrix} \begin{bmatrix} | & | & \dots & | \\ b_1 & b_2 & \dots & b_p \\ | & | & & | \end{bmatrix} = \begin{bmatrix} a_1^T b_1 & a_1^T b_2 & \dots & a_1^T b_p \\ a_2^T b_1 & a_2^T b_2 & \dots & a_2^T b_p \\ \vdots & \vdots & \ddots & \vdots \\ a_m^T b_1 & a_m^T b_2 & \dots & a_m^T b_p \end{bmatrix}$$

2. As a sum of outer products

$$C = AB = \begin{bmatrix} | & | & \dots & | \\ a_1 & a_2 & \dots & a_n \\ | & | & & | \end{bmatrix} \begin{bmatrix} - & b_1^T & - \\ - & b_2^T & - \\ & \vdots & \\ - & b_n^T & - \end{bmatrix} = \sum_{i=1}^n a_i b_i^T$$