

A Matrix's Rank and Calculating Inverse Matrices

Def. If $A \in M_{m \times n}(\mathbb{R})$, we define the **rank of A** , denoted $\text{Rank}(A)$, to be the rank of the linear transformation associated with A , L_A , $L_A: \mathbb{R}^n \rightarrow \mathbb{R}^m$.

If $m = n$, then notice that an $n \times n$ matrix is invertible if and only if its rank is n . This follows from an earlier theorem about linear transformations. This is because any matrix A is the matrix representation of a linear transformation. In fact we have:

Theorem: Let $T: V \rightarrow W$ be a linear transformation between finite dimensional vector spaces, and let B_1 and B_2 be ordered bases for V and W respectively. Then $\text{Rank}(T) = \text{Rank}([T]_{B_1}^{B_2})$.

Theorem: Let A be an $m \times n$ matrix. If P is an $m \times m$ matrix and Q is an $n \times n$ matrix, both invertible, then

- a. $\text{Rank}(AQ) = \text{Rank}(A)$
- b. $\text{Rank}(PA) = \text{Rank}(A)$
- c. $\text{Rank}(PAQ) = \text{Rank}(A)$.

Proof: a.
$$\begin{aligned} R(L_{AQ}) &= R(L_A L_Q) = L_A L_Q(\mathbb{R}^n) \\ &= L_A(L_Q(\mathbb{R}^n)) = L_A(\mathbb{R}^n) && \text{(since } L_Q \text{ is onto)} \\ &= R(L_A). \end{aligned}$$

Thus we have:

$$\text{Rank}(AQ) = \dim(R(L_{AQ})) = \dim(R(L_A)) = \text{Rank}(A).$$

b. Similarly, we have:

$$R(L_{PA}) = R(L_P(L_A)) = L_P(L_A(\mathbb{R}^n)).$$

But $L_A(\mathbb{R}^n)$ is a subspace of R^m .

Since P is invertible we have:

$$\dim(L_P(L_A(\mathbb{R}^n))) = \dim(L_A(\mathbb{R}^n)) = \text{Rank}(A).$$

So the $\text{Rank}(PA) = \text{Rank}(A)$.

c. Follows from parts a and b.

Corollary: Elementary row and column operations on a matrix are rank preserving.

Proof: Every elementary row or column operation can be viewed as a multiplication of a matrix by an invertible matrix on the left (elementary row operations) or the right (elementary column operations).

Theorem: The rank of any matrix equals the maximum number of its linearly independent columns. Thus the rank of a matrix is the dimension of the subspace generated by its columns.

Proof: Let $A \in M_{m \times n}(\mathbb{R})$.

$$\text{Rank}(A) = \text{Rank}(L_A) = \dim(R(L_A)).$$

Let B be the standard ordered basis for \mathbb{R}^n .

Then we have:

$$\begin{aligned} R(L_A) &= \text{span}\{L_A(B)\} \\ &= \text{span}\{L_A(e_1), \dots, L_A(e_n)\} \end{aligned}$$

But $L_A(e_j) = j^{\text{th}}$ column of $A = v_j$.

Thus $R(L_A) = \text{span}\{v_1, \dots, v_n\}$.

Hence $\text{Rank}(A) = \dim(R(L_A)) = \dim(\text{span}\{v_1, \dots, v_n\})$.

Ex. Let $A = \begin{bmatrix} 1 & 0 & 1 \\ 1 & 1 & 2 \\ 1 & 0 & 1 \end{bmatrix}$. Find the $\text{Rank}(A)$.

Notice that columns one and two are linearly independent, but column 3 is the sum of columns one and two. Thus we have:

$$\text{Rank}(A) = \dim\left(\text{Span}\left\{\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}\right\}\right) = 2.$$

In general it can be difficult to identify the maximum number of linearly independent columns of a matrix A . However, we know that we don't change $\text{Rank}(L_A)$ by performing elementary row (or column) operations (since they are invertible). Thus we can find the maximum number of linearly independent columns of a matrix A through elementary row and column operations.

Ex. Let $A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 1 & 5 \\ 2 & 3 & 3 \end{bmatrix}$. Find $\text{Rank}(A)$.

Subtracting 2(row 1) from row 2 and replacing it in row 2: $R_2 - 2R_1 \rightarrow R_2$.

$$\begin{bmatrix} 1 & 2 & 1 \\ 2 & 1 & 5 \\ 2 & 3 & 3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 1 \\ 0 & -3 & 3 \\ 2 & 3 & 3 \end{bmatrix}.$$

Now subtract 2(row 1) from row 3 and replace it in row 3: $R_3 - 2R_1 \rightarrow R_3$.

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -3 & 3 \\ 2 & 3 & 3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & 1 \\ 0 & -3 & 3 \\ 0 & -1 & 1 \end{bmatrix}.$$

Next subtract 2(column 1) from column 2 and replace it in column 2:
 $C_2 - 2C_1 \rightarrow C_2$.

$$\begin{bmatrix} 1 & 2 & 1 \\ 0 & -3 & 3 \\ 0 & -1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & -3 & 3 \\ 0 & -1 & 1 \end{bmatrix}.$$

Finally, subtract column 1 from column 3 and replace it in column 3: $C_3 - C_1 \rightarrow C_3$

$$\begin{bmatrix} 1 & 0 & 1 \\ 0 & -3 & 3 \\ 0 & -1 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & -3 & 3 \\ 0 & -1 & 1 \end{bmatrix}.$$

It's now clear that $C_3 = -C_2$ and the C_1 and C_2 are linearly independent. Thus $\text{Rank}(A) = 2$.

In fact, given an $m \times n$ matrix A we can always transform it using elementary row and column operations into a matrix that looks like:

$$D = \begin{bmatrix} I_r & \mathbf{0}_1 \\ \mathbf{0}_2 & \mathbf{0}_3 \end{bmatrix}$$

where $\mathbf{0}_1, \mathbf{0}_2$, and $\mathbf{0}_3$ are zero matrices.

Ex. Put $A = \begin{bmatrix} 0 & 2 & 2 & 0 \\ 2 & 2 & 0 & 4 \\ 4 & 6 & 4 & 6 \\ 0 & 4 & 2 & 2 \end{bmatrix}$ in the form $D = \begin{bmatrix} I_r & \mathbf{0}_1 \\ \mathbf{0}_2 & \mathbf{0}_3 \end{bmatrix}$ through elementary row

and column operations and find the $\text{Rank}(A)$.

$$\begin{bmatrix} 0 & 2 & 2 & 0 \\ 2 & 2 & 0 & 4 \\ 4 & 6 & 4 & 6 \\ 0 & 4 & 2 & 2 \end{bmatrix} \xrightarrow{R_2 \leftrightarrow R_1} \begin{bmatrix} 2 & 2 & 0 & 4 \\ 0 & 2 & 2 & 0 \\ 4 & 6 & 4 & 6 \\ 0 & 4 & 2 & 2 \end{bmatrix} \xrightarrow{\frac{1}{2}R_1 \rightarrow R_1} \begin{bmatrix} 1 & 1 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 4 & 6 & 4 & 6 \\ 0 & 4 & 2 & 2 \end{bmatrix}$$

$$\xrightarrow{R_3 - 4R_1 \rightarrow R_3} \begin{bmatrix} 1 & 1 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 0 & 2 & 4 & -2 \\ 0 & 4 & 2 & 2 \end{bmatrix} \xrightarrow{C_2 - C_1 \rightarrow C_2} \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 0 & 2 & 4 & -2 \\ 0 & 4 & 2 & 2 \end{bmatrix}$$

$$\xrightarrow{R_3 - R_2 \rightarrow R_3} \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 0 & 0 & 2 & -2 \\ 0 & 4 & 2 & 2 \end{bmatrix} \xrightarrow{R_4 - 2R_2 \rightarrow R_4} \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 0 & 0 & 2 & -2 \\ 0 & 0 & -2 & 2 \end{bmatrix}$$

$$\xrightarrow{R_3+R_4 \rightarrow R_4} \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 0 & 0 & 2 & -2 \\ 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{\frac{1}{2}R_3 \rightarrow R_3} \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 2 & 2 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\xrightarrow{\frac{1}{2}R_2 \rightarrow R_2} \begin{bmatrix} 1 & 0 & 0 & 2 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{C_4 - 2C_1 \rightarrow C_4} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$\xrightarrow{C_3 - C_2 \rightarrow C_3} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 \end{bmatrix} \xrightarrow{C_3 + C_4 \rightarrow C_4} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}.$$

Thus the $\text{Rank}(A) = 3$.

Note: One does not necessarily need to transform a matrix into the form

$$D = \begin{bmatrix} I_r & 0_1 \\ 0_2 & 0_3 \end{bmatrix} \text{ to identify its rank.}$$

Ex. Find the rank of $A = \begin{bmatrix} 1 & 3 & 1 & 1 \\ 2 & 0 & 2 & 2 \\ 1 & 6 & 1 & 1 \end{bmatrix}$.

$$\begin{bmatrix} 1 & 3 & 1 & 1 \\ 2 & 0 & 2 & 2 \\ 1 & 6 & 1 & 1 \end{bmatrix} \xrightarrow{R_2 - 2R_1 \rightarrow R_2} \begin{bmatrix} 1 & 3 & 1 & 1 \\ 0 & -6 & 0 & 0 \\ 1 & 6 & 1 & 1 \end{bmatrix} \xrightarrow{R_3 - R_1 \rightarrow R_3} \begin{bmatrix} 1 & 3 & 1 & 1 \\ 0 & -6 & 0 & 0 \\ 0 & 3 & 0 & 0 \end{bmatrix}.$$

It's now clear that there are two linearly independent column vectors so $\text{Rank}(A) = 2$.

Ex. Find the rank of $A = \begin{bmatrix} 1 & 2 & 3 & 1 \\ 2 & 1 & 1 & 1 \\ 1 & -1 & 1 & 0 \end{bmatrix}$.

$$\begin{bmatrix} 1 & 2 & 3 & 1 \\ 2 & 1 & 1 & 1 \\ 1 & -1 & 1 & 0 \end{bmatrix} \xrightarrow{R_2 - 2R_1 \rightarrow R_2} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -5 & -1 \\ 1 & -1 & 1 & 0 \end{bmatrix} \xrightarrow{R_3 - R_1 \rightarrow R_3} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -5 & -1 \\ 0 & -3 & -2 & -1 \end{bmatrix}$$

$$\xrightarrow{R_3 - R_2 \rightarrow R_3} \begin{bmatrix} 1 & 2 & 3 & 1 \\ 0 & -3 & -5 & -1 \\ 0 & 0 & 3 & 0 \end{bmatrix}.$$

It is now clear that $\langle 1, 0, 0 \rangle$, $\langle 2, -3, 0 \rangle$ and $\langle 3, -5, 3 \rangle$ are linearly independent in \mathbb{R}^3 . Since one can have at most 3 linearly independent vectors in \mathbb{R}^3 , the $\text{Rank}(A) = 3$.

Given any $n \times n$ matrix A we can put it in the form:

$$D = \begin{bmatrix} I_r & 0_1 \\ 0_2 & 0_3 \end{bmatrix}$$

using elementary row and column operations. In particular, if A is invertible then

$$D = \begin{bmatrix} 1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & 1 \end{bmatrix} = I_n.$$

So if A is $n \times n$ and invertible then there exist invertible matrices B and C such that:

$$I_n = BAC$$

where $B = E_p \cdots E_1$ and $C = G_1 \cdots G_q$ are products of elementary matrices.

But if we have $I_n = BAC$ then we have:

$$I_n = BAC$$

$$I_n C^{-1} = BACC^{-1}$$

$$C^{-1} = BA$$

$$CC^{-1} = CBA$$

$$I_n = CBA.$$

Thus we can write:

$$E_1 \cdots E_k A = I_n$$

where the E_i 's are elementary matrices.

But then we have:

$$E_1 \cdots E_k AA^{-1} = I_n A^{-1} = A^{-1}$$

$$E_1 \cdots E_k I_n = A^{-1}.$$

Thus to find A^{-1} we just need to apply to I_n the same elementary row operations that turned A into I_n .

Ex. Let $A = \begin{bmatrix} 1 & 4 & 3 \\ -1 & -2 & 0 \\ 2 & 2 & 3 \end{bmatrix}$. Find A^{-1} .

We start by creating the augmented matrix $(A|I_3)$.

We will then apply a sequence of elementary row operations that transform A into I_3 , to both A and I_3 .

	<u>Row Operations</u>
$\left[\begin{array}{ccc ccc} 1 & 4 & 3 & 1 & 0 & 0 \\ -1 & -2 & 0 & 0 & 1 & 0 \\ 2 & 2 & 3 & 0 & 0 & 1 \end{array} \right] \rightarrow \left[\begin{array}{ccc ccc} 1 & 4 & 3 & 1 & 0 & 0 \\ -1 & -2 & 0 & 0 & 1 & 0 \\ 0 & -6 & -3 & -2 & 0 & 1 \end{array} \right]$	$R_3 - 2R_1 \rightarrow R_3$
$\rightarrow \left[\begin{array}{ccc ccc} 1 & 4 & 3 & 1 & 0 & 0 \\ 0 & 2 & 3 & 1 & 1 & 0 \\ 0 & -6 & -3 & -2 & 0 & 1 \end{array} \right]$	$R_1 + R_2 \rightarrow R_2$
$\rightarrow \left[\begin{array}{ccc ccc} 1 & 4 & 3 & 1 & 0 & 0 \\ 0 & 2 & 3 & 1 & 1 & 0 \\ 0 & 0 & 6 & 1 & 3 & 1 \end{array} \right]$	$R_3 + 3R_2 \rightarrow R_3$
$\rightarrow \left[\begin{array}{ccc ccc} 1 & 2 & 0 & 0 & -1 & 0 \\ 0 & 2 & 3 & 1 & 1 & 0 \\ 0 & 0 & 6 & 1 & 3 & 1 \end{array} \right]$	$R_1 - R_2 \rightarrow R_1$
$\rightarrow \left[\begin{array}{ccc ccc} 1 & 2 & 0 & 0 & -1 & 0 \\ 0 & 2 & 0 & \frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & 6 & 1 & 3 & 1 \end{array} \right]$	$R_2 - \frac{1}{2}R_3 \rightarrow R_2$
$\rightarrow \left[\begin{array}{ccc ccc} 1 & 0 & 0 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ 0 & 2 & 0 & \frac{1}{2} & -\frac{1}{2} & -\frac{1}{2} \\ 0 & 0 & 6 & 1 & 3 & 1 \end{array} \right]$	$R_1 - R_2 \rightarrow R_1$
$\rightarrow \left[\begin{array}{ccc ccc} 1 & 0 & 0 & -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ 0 & 1 & 0 & \frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} \\ 0 & 0 & 1 & \frac{1}{6} & \frac{1}{2} & \frac{1}{6} \end{array} \right]$	$\frac{1}{6}R_3 \rightarrow R_3, \frac{1}{2}R_2 \rightarrow R_2$

$$\text{So } A^{-1} = \begin{bmatrix} -\frac{1}{2} & -\frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} \\ \frac{1}{6} & \frac{1}{2} & \frac{1}{6} \end{bmatrix}. \quad \text{A straight forward calculation will show}$$

that $AA^{-1} = A^{-1}A = I_3$.