

3.8 Rank of a Matrix

Definition 3.8.1. Let A be an $m \times n$ matrix with entries from a field F . Define:

- The *Row Space of A* ($RSp(A)$) as the span of the rows of A . This is a subspace of F^n .
- The *Row Rank of A* is $\dim(RSp(A))$.
- The *Column Space of A* ($CSp(A)$) as the span of the columns of A . This is a subspace of F^m .
- The *Column Rank of A* is $\dim(CSp(A))$.

Example 3.8.2. Let $F = \mathbb{R}$ and $A = \begin{pmatrix} 3 & 1 & 2 \\ 0 & -1 & 1 \end{pmatrix}$. Then,

$$RSp(A) = \text{Span}\{(3 \ 1 \ 2), (0 \ -1 \ 1)\},$$

$$CSp(A) = \text{Span}\left\{\begin{pmatrix} 3 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \end{pmatrix}\right\}.$$

Now the row vectors $(3 \ 1 \ 2)$ and $(0 \ -1 \ 1)$ are linearly independent so $\dim(RSp(A)) = 2$, so the column rank is 2. The set

$$\left\{\begin{pmatrix} 3 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \end{pmatrix}\right\}$$

is linearly dependent as

$$\begin{pmatrix} 3 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ -1 \end{pmatrix} + \begin{pmatrix} 2 \\ 1 \end{pmatrix}.$$

So

$$CSp(A) = \text{Span}\left\{\begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 2 \\ 1 \end{pmatrix}\right\},$$

which is linearly independent, so $\dim(CSp(A)) = 2$.

Procedure 3.8.3.

Calculating the row rank of a matrix A .

- Step 1: Reduce A to row echelon form using row operations:

$$A_{ech} = \begin{pmatrix} 1 & * & * & * & * & \dots \\ 0 & 0 & 1 & * & * & \dots \\ 0 & 0 & 0 & 1 & * & * \dots \\ \vdots & & & & & \\ 0 & \dots & & & & \end{pmatrix}$$

(Actually it doesn't matter whether the leading entries in each row are 1s or not.)

- Step 2: The row rank of A is the number of non-zero rows in A_{ech} . In fact it the non-zero

rows of A_{ech} form a basis for $RSp(A)$.

Justification

It will be enough to show:

Example 3.8.4. Find the row rank of $A = \begin{pmatrix} 1 & 2 & 5 \\ 2 & 1 & 0 \\ -1 & 4 & 15 \end{pmatrix}$

Example 3.8.5. Find the dimension of

$$W = \text{Span}\{(-1 \ 1 \ 0 \ 1), (2 \ 3 \ 1 \ 0), (0 \ 1 \ 2 \ 3)\} \subseteq \mathbb{R}^4.$$

We can find the column rank of a matrix in a very similar way to finding the row rank of a matrix.

Procedure 3.8.6. The columns of A are the rows of A^T so we can apply Procedure 3.8.3 to A^T .

Alternatively: use column operations to reduce A to “column echelon form and then count the non-zero columns.

Example 3.8.7. Let $A = \begin{pmatrix} 1 & 2 & 5 \\ 2 & 1 & 0 \\ -1 & 4 & 15 \end{pmatrix}$. Find the column rank of A . This equals the row rank of A^T .

Theorem 3.8.8. For any matrix A the row rank of A is equal to the column rank of A .

Example 3.8.9. Let $A = \begin{pmatrix} 1 & 2 & -1 & 0 \\ -1 & 1 & 0 & 1 \\ 0 & 3 & -1 & 1 \end{pmatrix}$

Note that $r_3 = r_1 + r_2$, so a basis for $RSp(A)$ is

$$\{\underbrace{(1, 2, -1, 0)}_{v_1}, \underbrace{(-1, 1, 0, 1)}_{v_2}\}$$

Write the rows as linear combinations of v_1 and v_2 :

$$r_1 =$$

$$r_2 =$$

$$r_3 =$$

These co-efficients are the λ_{ij} 's from the proof:

$$\begin{array}{ll} \lambda_{11} = & \lambda_{12} = \\ \lambda_{21} = & \lambda_{22} = \\ \lambda_{31} = & \lambda_{32} = \end{array}$$

According to the proof, a spanning set for $CSp(A)$ is:

$$\begin{pmatrix} \lambda_{11} \\ \lambda_{21} \\ \lambda_{31} \end{pmatrix} = \quad , \quad \begin{pmatrix} \lambda_{11} \\ \lambda_{21} \\ \lambda_{31} \end{pmatrix} =$$

Check this is really a spanning set for $CSP(A)$: Let $w_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}$, $w_2 = \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix}$

Now we have:

$$\begin{aligned} c_1 &= \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} = \\ c_2 &= \begin{pmatrix} 2 \\ 1 \\ 3 \end{pmatrix} = \\ c_3 &= \begin{pmatrix} -1 \\ 0 \\ -1 \end{pmatrix} = \\ c_4 &= \begin{pmatrix} 0 \\ 1 \\ 1 \end{pmatrix} = \end{aligned}$$

So it is indeed the case that $\{w_1, w_2\}$ spans $CSp(A)$.

Definition 3.8.10. Let A be a matrix. The *rank* of A written $\text{rank}(A)$ or $\text{rk}(A)$, is the row rank of A (or the column rank since they are the same).

Proposition 3.8.11. Let A be an $n \times n$ matrix with entries in F , then the following statements are equivalent:

1. $\text{rank}(A) = n$ (“ A has full rank”).
2. The rows of A form a basis for F^n .
3. The columns of A form a basis for F^n .
4. A is invertible (so $\det(A) \neq 0$, etc.).

4 Linear Transformations

4.1 Introduction

Definition 4.1.1. Suppose V, W are vector spaces over a field F . Let $T : V \rightarrow W$ be a function from V to W . We say:

- T preserves addition if for all $v_1, v_2 \in V$ we have $T(v_1 + v_2) = T(v_1) + T(v_2)$. (i.e. if $T(v_1) = w_1, T(v_2) = w_2$ for $w_1, w_2 \in W$ we have $T(v_1 + v_2) = w_1 + w_2$.)
- T preserves scalar multiplication if for all $v \in V, \lambda \in F, T(\lambda v) = \lambda T(v)$.
- T is a *linear transformation* (or *linear map*) if it:
 1. preserves addition.
 2. preserves scalar multiplication

Example 4.1.2.

- (a) The identity map $T : V \rightarrow V$ is obviously a linear transformation.
- (b) $T : \mathbb{R}^2 \rightarrow \mathbb{R}$ defined by $T(x, y) = x + y$ is a linear transformation.
- (c) Let V be the space of all polynomials in x over \mathbb{R} (i.e. $V = \mathbb{R}[x]$). Define $T : V \rightarrow V$ by $T(f(x)) = \frac{d}{dx}f(x)$. Then T is a linear map.
- (d) Let $V = \mathbb{C}$ (as a 1-dimensional vector space over \mathbb{C}). The map $T(z) = \bar{z}$ is **not** a linear map:
- (e) Let $T : \mathbb{R}^3 \rightarrow \mathbb{R}$ be given by $T(x, y, z) = (xyz)^{\frac{1}{3}}$ then:

- (f) Lots of functions preserve neither addition nor scalar multiplication, e.g., for $\mathbb{R} \rightarrow \mathbb{R}$ the functions taking $x \mapsto x + 1$, $x \mapsto x^2$, and $x \mapsto e^x$.

Proposition 4.1.3. Let A be an $m \times n$ matrix over F . Define $T : F^n \longrightarrow F^m$ (spaces of column vectors), by $T(v) = Av$ (for $v \in F^n$). Then T is a linear transformation.

Proposition 4.1.4. Basic Properties of linear transformations

Let $T : V \longrightarrow W$ be a linear map. Write $0_V, 0_W$ for the zero vectors in V and W respectively. We have:

1. $T(0_v) = 0_W$
2. Suppose $v = \lambda_1 v_1 + \dots + \lambda_k v_k$ for $\lambda_i \in F$, $v_i \in V$. Then $T(v) = \lambda_1 T(v_1) + \dots + \lambda_k T(v_k)$.