

April 1 Math 3260 sec. 52 Spring 2024

Section 4.5: Dimension of a Vector Space

Theorem:

If a vector space V has a basis $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$, then any set of vectors in V containing *more than n vectors* is linearly dependent.

This extends our result in \mathbb{R}^n that said that a set with more vectors than entries in each vector had to be linearly dependent.

For example,

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 10 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ 0 \\ 4 \end{bmatrix}, \begin{bmatrix} 3 \\ 6 \\ 8 \end{bmatrix} \right\}$$

is linearly dependent because there are 4 vectors from \mathbb{R}^3 .

Example

Recall that a basis for \mathbb{P}_3 is $\{1, t, t^2, t^3\}$.

Is the set below linearly dependent or linearly independent?

$$\{1 + t, 2t - 3t^3, 1 + t + t^2, 1 + t + t^2 + t^3, 2 - t + 2t^3\}$$

This set has 5 vectors. A basis has 4 vectors. Since $5 > 4$, this set is linearly dependent.

All Bases are the same Size

Our theorem gives the immediate corollary:

Corollary:

If vector space V has a basis $\mathcal{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_n\}$, then every basis of V consist of exactly n vectors.

Remark: This makes sense. If one basis had more vectors than another basis, it couldn't be linearly independent.

Dimension

Consider a vector space V .

Definition:

If V is spanned by a finite set, then V is called **finite dimensional**. In this case, the dimension of V

$\dim V =$ the number of vectors in any basis of V .

The dimension of the vector space $\{\mathbf{0}\}$ containing only the zero vector is defined to be zero—i.e.

$$\dim\{\mathbf{0}\} = 0.$$

If V is not spanned by a finite set^a, then V is said to be **infinite dimensional**.

^a $C^0(\mathbb{R})$ is an example of an infinite dimensional vector space.

Examples

(a) Determine $\dim(\mathbb{R}^n)$. = n

The standard basis has n vectors in it,
 $\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n$.

(b) Determine $\dim \text{Col}(A)$ where $A = \begin{bmatrix} 1 & 1 & 3 \\ 0 & 0 & -1 \end{bmatrix}$.

A is in ref. w/ two pivot columns.

Any basis will have two vectors

$$\text{so } \dim(\text{Col}(A)) = 2$$

Some Geometry in \mathbb{R}^3

We can describe all of the subspaces of \mathbb{R}^3 geometrically. The subspace(s) of dimension

- (a) **zero**: is just the origin (one point), $(0, 0, 0)$.
- (b) **one**: are lines through the origin. $\text{Span}\{\mathbf{u}\}$ where \mathbf{u} is not the zero vector.
- (c) **two**: are planes that contain the origin and two other, noncolinear points. $\text{Span}\{\mathbf{u}, \mathbf{v}\}$ with $\{\mathbf{u}, \mathbf{v}\}$ linearly independent.
- (d) **three**: is all of \mathbb{R}^3 .

Subspaces and Dimension

Theorem:

Let H be a subspace of a finite dimensional vector space V . Then H is finite dimensional and

$$\dim H \leq \dim V.$$

Moreover, any linearly independent subset of H can be expanded if needed to form a basis for H .

Remark: We said before that we can take a spanning set and remove extra vectors to get a basis. This follow up statement says if we start with a linearly independent set, we can add to it as needed to get a basis.

Subspaces and Dimension

Theorem:

Let V be a vector space with $\dim V = p$ where $p \geq 1$. Any linearly independent set in V containing exactly p vectors is a basis for V . Similarly, any spanning set consisting of exactly p vectors in V is necessarily a basis for V .

Remark: this connects two properties **spanning** and **linear independence**. If $\dim V = p$ and a set contains p vectors then

- ▶ linear independence \implies spanning
- ▶ spanning \implies linear independence

Again, this is **IF** the number of vectors matches the dimension of the vector space.

Column and Null Spaces

Theorem:

Let A be an $m \times n$ matrix. Then

$\dim \text{Nul } A =$ the number of free variables in the equation $A\mathbf{x} = \mathbf{0}$,

and

$\dim \text{Col } A =$ the number of pivot positions in A .

Example

A matrix A is shown along with its rref. Find the dimensions of the null space and column space of A .

$$A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ -3 & 1 & -7 & -1 \\ 3 & 0 & 6 & 1 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 2 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

A has 3 pivot columns and one nonpivot column.

$$\dim(\text{Col}(A)) = 3$$

There would be one free variable in $A\vec{x} = \vec{0}$

$$\text{Hence } \dim(\text{Nul}(A)) = 1.$$

Example

A matrix A along with its rref is shown.

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

(a) Find a basis for Row A and state \dim Row A .

A basis for Row(A) is

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -2 \\ 0 \\ 3 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 5 \end{bmatrix} \right\}.$$

$$\dim(\text{Row } A) = 3.$$

Example continued ...

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

(b) Find a basis for $\text{Col } A$ and state $\dim \text{Col } A$.

From the row echelon form, columns 1, 2, and 4 are pivot columns. A basis for $\text{Col } A$ is

$$\left\{ \begin{bmatrix} -2 \\ 1 \\ 3 \\ 1 \end{bmatrix}, \begin{bmatrix} -5 \\ 3 \\ 11 \\ 7 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 7 \\ 5 \end{bmatrix} \right\} \quad \dim(\text{Col } A) = 3$$

Example continued ...

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

(c) Find a basis for $\text{Nul } A$ and state $\dim \text{Nul } A$.

Note that if $A\vec{x} = \vec{0}$ then \vec{x} is in \mathbb{R}^5

$$x_1 = -x_3 - x_5$$

$$x_2 = 2x_3 - 3x_5$$

x_3 is free

$$x_4 = 5x_5$$

x_5 is free

$$\vec{x} = x_3 \begin{bmatrix} -1 \\ 2 \\ 1 \\ 0 \\ 0 \end{bmatrix} + x_5 \begin{bmatrix} -1 \\ -3 \\ 0 \\ 5 \\ 1 \end{bmatrix}$$

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

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

$$\dim(\text{Nul}(A)) = 2$$

Remarks

- ▶ Row operations preserve row space, but change linear dependence relations of rows.
- ▶ Row operations change column space, but preserve linear dependence relations of columns.
- ▶ Another way to obtain a basis for Row A is to take the transpose A^T and do row operations. We have the following relationships:

$$\text{Row } A = \text{Col } A^T \quad \text{and} \quad \text{Col } A = \text{Row } A^T$$

Rank & Nullity

Definition:

The **rank** of a matrix A , denoted $\text{rank}(A)$, is the dimension of the column space of A .

Definition:

The **nullity** of a matrix A is the dimension of the null space.

Remark: Since the dimension of the column space is the number of pivot positions, the dimensions of the column and row spaces are the same. That is,

$$\text{rank}(A) = \dim \text{Col}(A) = \dim \text{Row}(A).$$

The Rank-Nullity Theorem

Theorem:

For $m \times n$ matrix A , $\dim \text{Col}(A) = \dim \text{Row}(A) = \text{rank}(A)$. Moreover

$$\text{rank } A + \dim \text{Nul } A = n.$$

The Rank-Nullity Theorem

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$$\text{rank } A + \dim \text{Nul } A = n.$$

Note: This theorem states the rather obvious fact that

$$\left\{ \begin{array}{c} \text{number of} \\ \text{pivot columns} \end{array} \right\} + \left\{ \begin{array}{c} \text{number of} \\ \text{non-pivot columns} \end{array} \right\} = \left\{ \begin{array}{c} \text{total number} \\ \text{of columns} \end{array} \right\}.$$

Examples

$$\text{rank} + \text{nullity} = n.$$

(1) A is a 5×4 matrix and $\text{rank}(A) = 4$. What is $\dim \text{Nul } A$?

Here, $n=4$, and $\text{rank}(A)=4$.

$4 + \text{nullity} = 4 \Rightarrow$ the nullity of A is zero.

Note that $\text{Nul}(A) = \{ \vec{0} \}$ in \mathbb{R}^4 .

Does $A\vec{x} = \vec{0}$ has nontrivial solutions?

Examples

$$\text{rank} + \text{nullity} = n.$$

(2) Suppose A is 7×5 and $\dim \text{Col } A = 2$. Determine

1. the nullity of A $n=5$ and $\text{rank}(A) = \dim(\text{Col } A) = 2$

$$2 + \text{nullity} = 5 \Rightarrow \text{nullity is } 3.$$

2. the rank of A^T $\text{rank}(A^T) = \dim(\text{Col } A^T) = \dim(\text{Row } A)$
 $= \dim(\text{Col } A) = 2$

$$\text{rank}(A^T) = 2$$

3. the nullity of A^T for A^T , the "n" is 7.

$$2 + \text{nullity} = 7 \Rightarrow \text{nullity of } 5.$$

Addendum to Invertible Matrix Theorem

Theorem:

Let A be an $n \times n$ matrix. The following are equivalent to the statement that A is invertible.

- (m) The columns of A form a basis for \mathbb{R}^n
- (n) $\text{Col } A = \mathbb{R}^n$
- (o) $\dim \text{Col } A = n$
- (p) $\text{rank } A = n$
- (q) $\text{Nul } A = \{\mathbf{0}\}$
- (r) $\dim \text{Nul } A = 0$