

APPLIED LINEAR ALGEBRA

Chapter 3B. Subspaces, Span and Linear independence

Definition (Recall: Linear Combination)

Let $\vec{u}_1, \dots, \vec{u}_n, \vec{v}$ be vectors. Then \vec{v} is said to be a **linear combination** of the vectors $\vec{u}_1, \dots, \vec{u}_n$ if there exist scalars, a_1, \dots, a_n such that

$$\vec{v} = a_1\vec{u}_1 + \dots + a_n\vec{u}_n$$

Definition (Span of a Set of Vectors)

The collection of all linear combinations of a set of vectors $\{\vec{u}_1, \dots, \vec{u}_k\}$ in \mathbb{R}^n is known as the span of these vectors and is written as $\text{span}\{\vec{u}_1, \dots, \vec{u}_k\}$.

Additional Terminology. If $U = \text{span}\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$, then

- U is **spanned by** the vectors $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k$.
- the vectors $\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k$ **span** U .
- the set of vectors $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$ is a **spanning set** for U .

Problem

Let $\vec{u} = \begin{bmatrix} 1 & 1 & 0 \end{bmatrix}^T$ and $\vec{v} = \begin{bmatrix} 3 & 2 & 0 \end{bmatrix}^T \in \mathbb{R}^3$. Show that $\vec{w} = \begin{bmatrix} 4 & 5 & 0 \end{bmatrix}^T$ is in $\text{span}\{\vec{u}, \vec{v}\}$.

Solution

For a vector to be in $\text{span}\{\vec{u}, \vec{v}\}$, it must be a linear combination of these vectors. If $\vec{w} \in \text{span}\{\vec{u}, \vec{v}\}$, we must be able to find scalars a, b such that

$$\vec{w} = a\vec{u} + b\vec{v}$$

$$\begin{bmatrix} 4 \\ 5 \\ 0 \end{bmatrix} = a \begin{bmatrix} 1 \\ 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 3 \\ 2 \\ 0 \end{bmatrix}$$

This is equivalent to the following system of equations

$$a + 3b = 4$$

$$a + 2b = 5$$

Solution (continued)

We solve this system the usual way, constructing the augmented matrix and row reducing to find the reduced row-echelon form.

$$\left[\begin{array}{cc|c} 1 & 3 & 4 \\ 1 & 2 & 5 \end{array} \right] \rightarrow \cdots \rightarrow \left[\begin{array}{cc|c} 1 & 0 & 7 \\ 0 & 1 & -1 \end{array} \right]$$

The solution is $a = 7, b = -1$. This means that

$$\vec{w} = 7\vec{u} - \vec{v}$$

Therefore we can say that \vec{w} is in $\text{span}\{\vec{u}, \vec{v}\}$.

Span of a Set of Vectors

Example

Let $\vec{x} \in \mathbb{R}^3$ be a nonzero vector. Then $\text{span}\{\vec{x}\} = \{k\vec{x} \mid k \in \mathbb{R}\}$ is a line through the origin having direction vector \vec{x} .

Problem

Describe the span of the vectors $\vec{u} = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}$ and $\vec{v} = \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}$.

Solution

Notice that any linear combination of the vectors \vec{u} and \vec{v} yields a vector $\begin{bmatrix} 0 \\ y \\ z \end{bmatrix}$ in the YZ -plane.

Suppose we take an arbitrary vector $\begin{bmatrix} 0 \\ y \\ z \end{bmatrix}$ in the YZ -plane. It turns out we can write any such vector as a linear combination of \vec{u} and \vec{v} .

$$\begin{bmatrix} 0 \\ y \\ z \end{bmatrix} = (-3y + 2z) \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} + (2y - z) \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}$$

Hence, $\text{span}\{\vec{u}, \vec{v}\}$ is the YZ -plane.

Span of a Set of Vectors

Consider the previous example where the span of \vec{u} and \vec{v} was the YZ -plane. Suppose we add another vector \vec{w} , and consider the span of \vec{u} , \vec{v} , and \vec{w} . What would happen to the span?

Scenario 1 Suppose \vec{w} is a vector in the YZ -plane. For example,

$\vec{w} = \begin{bmatrix} 0 \\ 4 \\ 1 \end{bmatrix}$. Then \vec{w} is in the span of \vec{u} , \vec{v} . Adding \vec{w} to the set doesn't change the span at all.

$$\text{span} \{ \vec{u}, \vec{v}, \vec{w} \} = \text{span} \{ \vec{u}, \vec{v} \}$$

Span of a Set of Vectors

Scenario 2 Suppose \vec{w} is not in the YZ -plane. For example, suppose

$$\vec{w} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}.$$

Notice that now, the three vectors span \mathbb{R}^3 . Any vector in \mathbb{R}^3 can be written as a linear combination of \vec{u} , \vec{v} , \vec{w} as follows:

$$\begin{bmatrix} x \\ y \\ z \end{bmatrix} = (-4x + 5y + 2z) \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} + (2x + 2y - z) \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix} + (x) \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$

You can see that the span of these three vectors depended on whether \vec{w} was in $\text{span}\{\vec{u}, \vec{v}\}$ or not. In the next section, we will examine the distinction between these two scenarios using the concept of linear independence.

Linearly Independent Set of Vectors

Definition

Let $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$ be a set of vectors in \mathbb{R}^n . This set is **linearly independent** if no vector in the set is in the span of the other vectors of that set.

If a set of vectors is not linearly independent, we call it **linearly dependent**.

A Linearly Dependent Set

Problem

Consider the vectors $\vec{u} = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}$, $\vec{v} = \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}$, $\vec{w} = \begin{bmatrix} 0 \\ 4 \\ 1 \end{bmatrix}$. Is the set $\{\vec{u}, \vec{v}, \vec{w}\}$ linearly independent?

Solution

Notice that we can write \vec{w} as a linear combination of \vec{u}, \vec{v} as follows:

$$\begin{bmatrix} 0 \\ 4 \\ 1 \end{bmatrix} = (-10) \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} + (7) \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}$$

Hence, \vec{w} is in $\text{span}\{\vec{u}, \vec{v}\}$. By the definition, this set is not linearly independent (it is linearly dependent).

A Linearly Independent Set

Problem

Consider the vectors

$$\vec{u} = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}, \vec{v} = \begin{bmatrix} 0 \\ 2 \\ 3 \end{bmatrix}, \vec{w} = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix}$$

Is the set $\{\vec{u}, \vec{v}, \vec{w}\}$ linearly independent?

Solution

We cannot write any of the three vectors as a linear combination of the other two. (We will see how to show this soon.) Therefore the set $\{\vec{u}, \vec{v}, \vec{w}\}$ is **linearly independent**.

Linear Independence as a Linear Combination

The following theorem provides a familiar way to check if a set of vectors is linearly independent.

Theorem

The collection of vectors, $\{\vec{u}_1, \dots, \vec{u}_k\}$ in \mathbb{R}^n is linearly independent if and only if whenever

$$\sum_{i=1}^n a_i \vec{u}_i = \vec{0}$$

it follows that each $a_i = 0$.

Thus $\{\vec{u}_1, \dots, \vec{u}_k\}$ in \mathbb{R}^n is linearly independent exactly when the system of linear equations $AX = 0$ has only the trivial solution, where A is the $n \times k$ matrix having these vectors as columns.

Linear Independence

We can state the conclusion of this theorem in another way: The set of vectors $\{\vec{u}_1, \dots, \vec{u}_k\}$ is linearly independent if and only if there is no nontrivial linear combination which equals zero. If a linear combination of the vectors equals zero, then all the coefficients of the combination are zero.

If the set is linearly independent, then

$$a_1 \vec{u}_1 + \dots + a_k \vec{u}_k = 0$$

implies that

$$a_1 = a_2 = \dots = a_k = 0$$

Linear Independence

We can use the reduced row-echelon form of the matrix to determine if the columns form a linearly independent set of vectors.

Problem

Determine whether the following set of vectors are linearly independent.

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

Solution

Construct the 3×3 matrix A having these vectors as columns:

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

By the above theorem, the set of vectors is linearly independent if the system $AX = 0$ has only the trivial solution. We can see this from the reduced row-echelon form of the matrix A .

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

Since all columns are pivot columns (and the rank of A is 3), the vectors are linearly independent.

Problem

Determine whether the following vectors are linearly independent. If they are linearly dependent, write one of the vectors as a linear combination of the others.

$$\left\{ \begin{bmatrix} 1 \\ 2 \\ 4 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 7 \\ 17 \\ 2 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 3 \\ 0 \end{bmatrix}, \begin{bmatrix} 8 \\ 5 \\ 11 \\ 11 \end{bmatrix} \right\}$$

Solution

Construct the matrix A using these vectors as columns.

$$A = \begin{bmatrix} 1 & 2 & 0 & 8 \\ 2 & 7 & 1 & 5 \\ 4 & 17 & 3 & 11 \\ 1 & 2 & 0 & 11 \end{bmatrix}$$

Solution (continued)

The reduced row-echelon form of this matrix is

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

Since the rank of A is $3 < 4$, the vectors are linearly dependent.

Therefore, there are infinitely many solutions to $AX = 0$, one of which is

$$\begin{bmatrix} -2 \\ -1 \\ -3 \\ 1 \end{bmatrix}$$

Solution (continued)

Therefore we can write:

$$-2 \begin{bmatrix} 1 \\ 2 \\ 4 \\ 1 \end{bmatrix} - 1 \begin{bmatrix} 0 \\ 1 \\ 3 \\ 0 \end{bmatrix} - 3 \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix} + 1 \begin{bmatrix} 8 \\ 5 \\ 11 \\ 11 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

We can rewrite this as:

$$2 \begin{bmatrix} 1 \\ 2 \\ 4 \\ 1 \end{bmatrix} + 1 \begin{bmatrix} 0 \\ 1 \\ 3 \\ 0 \end{bmatrix} + 3 \begin{bmatrix} 2 \\ 0 \\ 0 \\ 3 \end{bmatrix} = \begin{bmatrix} 8 \\ 5 \\ 11 \\ 11 \end{bmatrix}$$

This shows that one of the vectors can be written as a linear combination of the other three vectors. While here we chose the fourth vector, we could have chosen any of the vectors to isolate.

Subspaces

Definition

Let V be a nonempty collection of vectors in \mathbb{R}^n . Then V is a **subspace** if whenever a and b are scalars and \vec{u} and \vec{v} are vectors in V , $a\vec{u} + b\vec{v}$ is also in V .

Subspaces are closely related to the span of a set of vectors which we discussed earlier.

Theorem

Let V be a nonempty collection of vectors in \mathbb{R}^n . Then V is a subspace of \mathbb{R}^n if and only if there exist vectors $\{\vec{u}_1, \dots, \vec{u}_k\}$ in V such that

$$V = \text{span} \{ \vec{u}_1, \dots, \vec{u}_k \}$$

Subspaces

Subspaces are also related to the property of linear independence.

Theorem

If V is a subspace of \mathbb{R}^n , then there exist linearly independent vectors $\{\vec{u}_1, \dots, \vec{u}_k\}$ of V such that

$$V = \text{span} \{ \vec{u}_1, \dots, \vec{u}_k \}$$

In other words, subspaces of \mathbb{R}^n consist of spans of finite, linearly independent collections of vectors in \mathbb{R}^n .

Basis of a Subspace

Definition

Let V be a subspace of \mathbb{R}^n . Then $\{\vec{u}_1, \dots, \vec{u}_k\}$ is called a **basis** for V if the following conditions hold:

- $\text{span}\{\vec{u}_1, \dots, \vec{u}_k\} = V$
- $\{\vec{u}_1, \dots, \vec{u}_k\}$ is linearly independent.

The following theorem claims that any two bases of a subspace must be of the same size.

Theorem

Let V be a subspace of \mathbb{R}^n and suppose $\{\vec{u}_1, \dots, \vec{u}_k\}$ and $\{\vec{v}_1, \dots, \vec{v}_m\}$ are two bases for V . Then $k = m$.

Dimension

The previous theorem shows that all bases of a subspace will have the same size. This size is called the **dimension** of the subspace.

Definition

Let V be a subspace of \mathbb{R}^n . Then the **dimension** of V is the number of vectors in a basis of V .

Properties of \mathbb{R}^n

Note that the dimension of \mathbb{R}^n is n .

There are some other important properties of vectors in \mathbb{R}^n .

Theorem

- If $\{\vec{u}_1, \dots, \vec{u}_n\}$ is a linearly independent set of n vectors in \mathbb{R}^n , then $\{\vec{u}_1, \dots, \vec{u}_n\}$ is a basis for \mathbb{R}^n .
- Suppose $\{\vec{u}_1, \dots, \vec{u}_m\}$ spans \mathbb{R}^n . Then $m \geq n$.
- If $\{\vec{u}_1, \dots, \vec{u}_n\}$ spans \mathbb{R}^n , then $\{\vec{u}_1, \dots, \vec{u}_n\}$ is linearly independent.
- If $\{\vec{u}_1, \vec{u}_2, \dots, \vec{u}_k\}$ is a set of vectors in \mathbb{R}^n with $k > n$, then the set is linearly dependent.

It follows then that a basis is a minimal spanning set. If a subspace has dimension d , then any spanning set has size at least d , and any spanning set of size d must be a basis (and is therefore independent).

Row and Column Space

Definition

Let A be an $m \times n$ matrix. The **column space** of A is the span of the columns of A . The **row space** of A is the span of the rows of A .

Problem

Find the rank of the matrix A and describe the column and row spaces efficiently.

$$A = \begin{bmatrix} 1 & 2 & 1 & 3 & 2 \\ 1 & 3 & 6 & 0 & 2 \\ 3 & 7 & 8 & 6 & 6 \end{bmatrix}$$

Example: Column Space

Solution

To find the column space, we first find the reduced row-echelon form of A :

$$\begin{bmatrix} 1 & 0 & -9 & 9 & 2 \\ 0 & 1 & 5 & -3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

Therefore $\text{rank}(A) = 2$.

Note the first two columns are the pivot columns. All columns of the above reduced row-echelon matrix are in

$$\text{span} \left\{ \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \right\}$$

Example: Column Space

Solution (continued)

To construct the column space, we use the pivot columns of the original matrix - in this case, the first and second columns. Therefore the column space of A is

$$\text{span} \left\{ \begin{bmatrix} 1 \\ 1 \\ 3 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 7 \end{bmatrix} \right\}$$

Example: Row Space

Solution (continued)

To find the row space of A we again look at the reduced row-echelon form of the matrix.

$$\begin{bmatrix} 1 & 0 & -9 & 9 & 2 \\ 0 & 1 & 5 & -3 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

The row space of A is the span of the non-zero rows of the above matrix:

$$\text{span} \left\{ \begin{bmatrix} 1 & 0 & -9 & 9 & 2 \end{bmatrix}, \begin{bmatrix} 0 & 1 & 5 & -3 & 0 \end{bmatrix} \right\}$$

Notice that the vectors used in the description of the **column space** are **from the original matrix**, while those in the **row space** are **from the reduced row-echelon form of the original matrix**.

Null Space

Definition

The **null space** of A , or **kernel** of A is defined as:

$$\ker(A) = \{X : AX = 0\}$$

We also speak of the image of A , $\text{Im}(A)$, which is all vectors of the form AX where X is in \mathbb{R}^n .

To find $\ker(A)$, we solve the system of equations $AX = 0$.

Problem

Find $\ker(A)$ for the matrix A :

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

Null Space

Solution

The first step is to set up the augmented matrix:

$$\left[\begin{array}{ccc|c} 1 & 2 & 1 & 0 \\ 0 & -1 & 1 & 0 \\ 2 & 3 & 3 & 0 \end{array} \right]$$

Place this matrix in reduced row-echelon form:

$$\left[\begin{array}{ccc|c} 1 & 0 & 3 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

Null Space

Solution (continued)

The solution to this system of equations is

$$\left\{ \begin{bmatrix} 3t \\ t \\ t \end{bmatrix} : t \in \mathbb{R} \right\}$$

Therefore the null space of A is the span of this vector:

$$\ker(A) = \text{span} \left\{ \begin{bmatrix} 3 \\ 1 \\ 1 \end{bmatrix} \right\}$$

Nullity

Definition

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

Theorem

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

$$\text{rank}(A) + \text{null}(A) = n$$

For instance, in the last example, A was a 3×3 matrix. The rank was 2 and the nullity was 1 (since the null space had dimension 1).

$$\text{rank}(A) + \text{null}(A) = 2 + 1 = 3 = n$$