MATH 323

Linear Algebra

Lecture 10:

Span. Spanning set. Linear independence.

Subspaces of vector spaces

Definition. A vector space V_0 is a **subspace** of a vector space V if $V_0 \subset V$ and the linear operations on V_0 agree with the linear operations on V.

Proposition A subset S of a vector space V is a subspace of V if and only if S is **nonempty** and **closed under linear operations**, i.e.,

$$\mathbf{x}, \mathbf{y} \in S \implies \mathbf{x} + \mathbf{y} \in S,$$

 $\mathbf{x} \in S \implies r\mathbf{x} \in S \text{ for all } r \in \mathbb{R}.$

Remarks. The zero vector in a subspace is the same as the zero vector in V. Also, the subtraction in a subspace agrees with that in V.

Let V be a vector space and $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n \in V$. Consider the set L of all linear combinations $r_1\mathbf{v}_1 + r_2\mathbf{v}_2 + \dots + r_n\mathbf{v}_n$, where $r_1, r_2, \dots, r_n \in \mathbb{R}$.

Theorem L is a subspace of V.

Proof: First of all, L is not empty. For example, $\mathbf{0} = 0\mathbf{v}_1 + 0\mathbf{v}_2 + \cdots + 0\mathbf{v}_n$ belongs to L.

The set L is closed under addition since

$$(r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_n\mathbf{v}_n)+(s_1\mathbf{v}_1+s_2\mathbf{v}_2+\cdots+s_n\mathbf{v}_n)=$$

= $(r_1+s_1)\mathbf{v}_1+(r_2+s_2)\mathbf{v}_2+\cdots+(r_n+s_n)\mathbf{v}_n.$

The set L is closed under scalar multiplication since $t(r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_n\mathbf{v}_n)=(tr_1)\mathbf{v}_1+(tr_2)\mathbf{v}_2+\cdots+(tr_n)\mathbf{v}_n.$

Thus L is a subspace of V.

Span: implicit definition

Let S be a subset of a vector space V.

Definition. The **span** of the set S, denoted Span(S), is the smallest subspace of V that contains S. That is,

- $\operatorname{Span}(S)$ is a subspace of V;
- for any subspace $W \subset V$ one has $S \subset W \implies \operatorname{Span}(S) \subset W$.

Remark. The span of any set $S \subset V$ is well defined (namely, it is the intersection of all subspaces of V that contain S).

Span: effective description

Let S be a subset of a vector space V.

- If $S = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ then $\mathrm{Span}(S)$ is the set of all linear combinations $r_1\mathbf{v}_1 + r_2\mathbf{v}_2 + \dots + r_n\mathbf{v}_n$, where $r_1, r_2, \dots, r_n \in \mathbb{R}$.
- If S is an infinite set then $\mathrm{Span}(S)$ is the set of all linear combinations $r_1\mathbf{u}_1+r_2\mathbf{u}_2+\cdots+r_k\mathbf{u}_k$, where $\mathbf{u}_1,\mathbf{u}_2,\ldots,\mathbf{u}_k\in S$ and $r_1,r_2,\ldots,r_k\in\mathbb{R}$ $(k\geq 1)$.
 - If S is the empty set then $\operatorname{Span}(S) = \{\mathbf{0}\}.$

Examples of subspaces of $\mathcal{M}_{2,2}(\mathbb{R})$:

• The span of $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$ and $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$ consists of all

• The span of
$$\begin{pmatrix} 0 & 0 \end{pmatrix}$$
 and $\begin{pmatrix} 0 & 1 \end{pmatrix}$ consists of a matrices of the form
$$a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} a & 0 \\ 0 & b \end{pmatrix}.$$

This is the subspace of diagonal matrices.

• The span of
$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$
, $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$, and $\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ consists of all matrices of the form
$$a\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix} + c\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} a & c \\ c & b \end{pmatrix}.$$

This is the subspace of symmetric matrices $(A^T = A)$.

Examples of subspaces of $\mathcal{M}_{2,2}(\mathbb{R})$:

- The span of $\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$ is the subspace of anti-symmetric matrices $(A^T = -A)$.
- The span of $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$, $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$, and $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$ is the subspace of upper triangular matrices.
- The span of $\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$, $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$, $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$, $\begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$ is the entire space $\mathcal{M}_{2,2}(\mathbb{R})$.

Spanning set

Definition. A subset S of a vector space V is called a **spanning set** for V if Span(S) = V.

Examples.

• Vectors $\mathbf{e}_1 = (1, 0, 0)$, $\mathbf{e}_2 = (0, 1, 0)$, and $\mathbf{e}_3 = (0, 0, 1)$ form a spanning set for \mathbb{R}^3 as $(x, y, z) = x\mathbf{e}_1 + y\mathbf{e}_2 + z\mathbf{e}_3$.

• Matrices
$$\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}$$
, $\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix}$, $\begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}$, $\begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}$

form a spanning set for $\mathcal{M}_{2,2}(\mathbb{R})$ as

$$\begin{pmatrix} a & b \\ c & d \end{pmatrix} = a \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix} + b \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} + c \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} + d \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}.$$

Problem Let $\mathbf{v}_1 = (1, 2, 0)$, $\mathbf{v}_2 = (3, 1, 1)$, and $\mathbf{w} = (4, -7, 3)$. Determine whether \mathbf{w} belongs to $\mathrm{Span}(\mathbf{v}_1, \mathbf{v}_2)$.

We have to check if there exist $r_1, r_2 \in \mathbb{R}$ such that $\mathbf{w} = r_1 \mathbf{v}_1 + r_2 \mathbf{v}_2$. This vector equation is equivalent to a system of linear equations:

$$\begin{pmatrix} 4 \\ -7 \\ 3 \end{pmatrix} = r_1 \begin{pmatrix} 1 \\ 2 \\ 0 \end{pmatrix} + r_2 \begin{pmatrix} 3 \\ 1 \\ 1 \end{pmatrix} \iff \begin{cases} 4 = r_1 + 3r_2 \\ -7 = 2r_1 + r_2 \\ 3 = 0r_1 + r_2 \end{cases}$$

The system has a unique solution: $r_1 = -5$, $r_2 = 3$. Thus $\mathbf{w} = -5\mathbf{v}_1 + 3\mathbf{v}_2$ is in $\mathrm{Span}(\mathbf{v}_1, \mathbf{v}_2)$. **Problem** Let $\mathbf{v}_1 = (2,5)$ and $\mathbf{v}_2 = (1,3)$. Show that $\{\mathbf{v}_1, \mathbf{v}_2\}$ is a spanning set for \mathbb{R}^2 .

Take any vector $\mathbf{w} = (a, b) \in \mathbb{R}^2$. We have to check that there exist $r_1, r_2 \in \mathbb{R}$ such that

$$\mathbf{w} = r_1 \mathbf{v}_1 + r_2 \mathbf{v}_2 \iff \begin{cases} 2r_1 + r_2 = a \\ 5r_1 + 3r_2 = b \end{cases}$$

Coefficient matrix: $C = \begin{pmatrix} 2 & 1 \\ 5 & 3 \end{pmatrix}$. det $C = 1 \neq 0$.

Since the matrix C is invertible, the system has a unique solution for any a and b.

Thus $\operatorname{Span}(\mathbf{v}_1,\mathbf{v}_2)=\mathbb{R}^2$.

Problem Let $\mathbf{v}_1 = (2,5)$ and $\mathbf{v}_2 = (1,3)$. Show that $\{\mathbf{v}_1, \mathbf{v}_2\}$ is a spanning set for \mathbb{R}^2 .

Alternative solution: First let us show that vectors $\mathbf{e}_1 = (1,0)$ and $\mathbf{e}_2 = (0,1)$ belong to $\mathrm{Span}(\mathbf{v}_1,\mathbf{v}_2)$.

$$\mathbf{e}_1 = r_1 \mathbf{v}_1 + r_2 \mathbf{v}_2 \iff \begin{cases} 2r_1 + r_2 = 1 \\ 5r_1 + 3r_2 = 0 \end{cases} \iff \begin{cases} r_1 = 3 \\ r_2 = -5 \end{cases}$$

$$\mathbf{e}_2 = r_1 \mathbf{v}_1 + r_2 \mathbf{v}_2 \iff \begin{cases} 2r_1 + r_2 = 0 \\ 5r_1 + 3r_2 = 1 \end{cases} \iff \begin{cases} r_1 = -1 \\ r_2 = 2 \end{cases}$$
Thus, $\mathbf{e}_1 = 3\mathbf{v}_1 = 5\mathbf{v}_2$ and $\mathbf{e}_2 = -\mathbf{v}_1 + 2\mathbf{v}_2$

Thus $\mathbf{e}_1 = 3\mathbf{v}_1 - 5\mathbf{v}_2$ and $\mathbf{e}_2 = -\mathbf{v}_1 + 2\mathbf{v}_2$. Then for any vector $\mathbf{w} = (a, b) \in \mathbb{R}^2$ we have $\mathbf{w} = a\mathbf{e}_1 + b\mathbf{e}_2 = a(3\mathbf{v}_1 - 5\mathbf{v}_2) + b(-\mathbf{v}_1 + 2\mathbf{v}_2) = (3a - b)\mathbf{v}_1 + (-5a + 2b)\mathbf{v}_2$. **Problem** Let $\mathbf{v}_1 = (2,5)$ and $\mathbf{v}_2 = (1,3)$. Show that $\{\mathbf{v}_1, \mathbf{v}_2\}$ is a spanning set for \mathbb{R}^2 .

Remarks on the alternative solution:

Notice that \mathbb{R}^2 is spanned by vectors $\mathbf{e}_1 = (1,0)$ and $\mathbf{e}_2 = (0,1)$ since $(a,b) = a\mathbf{e}_1 + b\mathbf{e}_2$.

This is why we have checked that vectors \mathbf{e}_1 and \mathbf{e}_2 belong to $\mathrm{Span}(\mathbf{v}_1,\mathbf{v}_2)$. Then

$$\mathbf{e}_1, \mathbf{e}_2 \in \operatorname{Span}(\mathbf{v}_1, \mathbf{v}_2) \implies \operatorname{Span}(\mathbf{e}_1, \mathbf{e}_2) \subset \operatorname{Span}(\mathbf{v}_1, \mathbf{v}_2)$$

 $\implies \mathbb{R}^2 \subset \operatorname{Span}(\mathbf{v}_1, \mathbf{v}_2) \implies \operatorname{Span}(\mathbf{v}_1, \mathbf{v}_2) = \mathbb{R}^2.$

In general, to show that $\operatorname{Span}(S_1) = \operatorname{Span}(S_2)$, it is enough to check that $S_1 \subset \operatorname{Span}(S_2)$ and $S_2 \subset \operatorname{Span}(S_1)$.

More properties of span

Let S_0 and S be subsets of a vector space V.

• $S_0 \subset S \implies \operatorname{Span}(S_0) \subset \operatorname{Span}(S)$.

 $\mathbf{v}_1, \dots, \mathbf{v}_k$ is also a spanning set for V.

- $\operatorname{Span}(S_0) = V$ and $S_0 \subset S \Longrightarrow \operatorname{Span}(S) = V$.
- If $\mathbf{v}_0, \mathbf{v}_1, \dots, \mathbf{v}_k$ is a spanning set for V and \mathbf{v}_0 is a linear combination of vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ then

Indeed, if $\mathbf{v}_0 = r_1 \mathbf{v}_1 + \cdots + r_k \mathbf{v}_k$, then $t_0 \mathbf{v}_0 + t_1 \mathbf{v}_1 + \cdots + t_k \mathbf{v}_k = (t_0 r_1 + t_1) \mathbf{v}_1 + \cdots + (t_0 r_k + t_k) \mathbf{v}_k$.

• $\operatorname{Span}(S_0 \cup \{\mathbf{v}_0\}) = \operatorname{Span}(S_0)$ if and only if $\mathbf{v}_0 \in \operatorname{Span}(S_0)$.

If $\mathbf{v}_0 \in \operatorname{Span}(S_0)$, then $S_0 \cup \{\mathbf{v}_0\} \subset \operatorname{Span}(S_0)$, which implies $\operatorname{Span}(S_0 \cup \{\mathbf{v}_0\}) \subset \operatorname{Span}(S_0)$. On the other hand, $\operatorname{Span}(S_0) \subset \operatorname{Span}(S_0 \cup \{\mathbf{v}_0\})$.

Linear independence

Definition. Let V be a vector space. Vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k \in V$ are called **linearly dependent** if they satisfy a relation

$$r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_k\mathbf{v}_k=\mathbf{0},$$

where the coefficients $r_1, \ldots, r_k \in \mathbb{R}$ are not all equal to zero. Otherwise vectors $\mathbf{v}_1, \mathbf{v}_2, \ldots, \mathbf{v}_k$ are called **linearly independent**. That is, if

$$r_1\mathbf{v}_1+r_2\mathbf{v}_2+\cdots+r_k\mathbf{v}_k=\mathbf{0} \implies r_1=\cdots=r_k=0.$$

A set $S \subset V$ is **linearly dependent** if one can find some distinct linearly dependent vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ in S. Otherwise S is **linearly independent**.

Examples of linear independence

• Vectors $\mathbf{e}_1 = (1,0,0)$, $\mathbf{e}_2 = (0,1,0)$, and $\mathbf{e}_3 = (0,0,1)$ in \mathbb{R}^3 .

$$x\mathbf{e}_1 + y\mathbf{e}_2 + z\mathbf{e}_3 = \mathbf{0} \implies (x, y, z) = \mathbf{0}$$

 $\implies x = y = z = 0$

 $\bullet \ \ \mathsf{Matrices} \ \ E_{11} = \begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}, \ E_{12} = \begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix},$ $E_{21} = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \ \mathsf{and} \ \ E_{22} = \begin{pmatrix} 0 & 0 \\ 0 & 1 \end{pmatrix}.$

$$aE_{11} + bE_{12} + cE_{21} + dE_{22} = O \implies \begin{pmatrix} a & b \\ c & d \end{pmatrix} = O$$

$$\implies a = b = c = d = 0$$

Examples of linear independence

• Polynomials $1, x, x^2, \dots, x^n$.

$$a_0 + a_1 x + a_2 x^2 + \dots + a_n x^n = 0$$
 identically $\implies a_i = 0$ for $0 \le i \le n$

- The infinite set $\{1, x, x^2, \dots, x^n, \dots\}$.
- Polynomials $p_1(x) = 1$, $p_2(x) = x 1$, and $p_3(x) = (x 1)^2$.

$$a_1p_1(x) + a_2p_2(x) + a_3p_3(x) = a_1 + a_2(x-1) + a_3(x-1)^2 = (a_1 - a_2 + a_3) + (a_2 - 2a_3)x + a_3x^2.$$

Hence
$$a_1p_1(x) + a_2p_2(x) + a_3p_3(x) = 0$$
 identically $\Rightarrow a_1 - a_2 + a_3 = a_2 - 2a_3 = a_3 = 0$

$$\implies a_1 = a_2 = a_3 = 0$$

Problem Let $\mathbf{v}_1 = (1, 2, 0)$, $\mathbf{v}_2 = (3, 1, 1)$, and $\mathbf{v}_3 = (4, -7, 3)$. Determine whether vectors $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ are linearly independent.

We have to check if there exist $r_1, r_2, r_3 \in \mathbb{R}$ not all zero such that $r_1\mathbf{v}_1 + r_2\mathbf{v}_2 + r_3\mathbf{v}_3 = \mathbf{0}$.

This vector equation is equivalent to a system

$$\begin{cases} r_1 + 3r_2 + 4r_3 = 0 \\ 2r_1 + r_2 - 7r_3 = 0 \\ 0r_1 + r_2 + 3r_3 = 0 \end{cases} \begin{pmatrix} 1 & 3 & 4 & 0 \\ 2 & 1 & -7 & 0 \\ 0 & 1 & 3 & 0 \end{pmatrix}$$

The vectors $\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3$ are linearly dependent if and only if the coefficient matrix $A = (\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3)$ is singular. We obtain that $\det A = 0$ (it is singular).