# Lecture 13 : Two Basic Prototypes for Subspaces

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#### Agenda

- Subspaces
- Counter-Examples of Subspaces
- Two Basic Prototypes for Subspaces
  - $\blacktriangleright span(\mathbf{w}_1,\ldots,\mathbf{w}_k) \equiv \{t_1\mathbf{w}_1+\cdots+\mathbf{w}_k \mid t_1,\ldots,t_k \in \mathbb{R}\}$
  - The Solution Set of a Homogeneous Linear System Ax = 0

Closure Under Scalar Multiplication and Vector Addition

#### Definition

A subset  $S \subseteq \mathbb{R}^n$  is closed under scalar multiplication if

$$\lambda \in \mathbb{R} \text{ and } \mathbf{v} \in S \implies (\lambda \mathbf{v}) \in S$$

A subset  $S \subseteq \mathbb{R}^n$  is closed under vector addition if

$$\mathbf{v}_1, \mathbf{v}_2 \in S \implies (\mathbf{v}_1 + \mathbf{v}_2) \in S$$

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# Subspaces

## Definition

A **subspace** of  $\mathbb{R}^n$  is a subset *S* of  $\mathbb{R}^n$  that is closed under both scalar multiplication and vector addition:

$$egin{array}{ll} \lambda \in \mathbb{R} ext{ and } \mathbf{v} \in S & \Longrightarrow & (\lambda \mathbf{v}) \in S \ \mathbf{v}_1, \mathbf{v}_2 \in S & \Longrightarrow & (\mathbf{v}_1 + \mathbf{v}_2) \in S \end{array}$$

As the nomenclature suggests, **subspaces** can be thought of as vector spaces sitting inside a larger vector space (like a subset is a smaller set sitting inside a larger set).

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# Example: the unit circle $S^1$

Let

$$S^1 = \{ [x, y] \in \mathbb{R}^2 \mid x^2 + y^2 = 1 \}$$

This set is not closed under scalar multiplication. E.g.,

$$\lambda = 2 \in \mathbb{R} , \ \mathbf{v} = [1,0] \in S^1 \quad \Rightarrow \quad \lambda \mathbf{v} = [2,0] \notin S^1$$

since  $2^2 + 0^2 \neq 1$  $S^1$  is **not closed** under vector addition either. E.g.,

$$\left[ 1,0 
ight], \left[ 0,1 
ight] \in S^{1} \quad \Rightarrow \quad \left[ 1,0 
ight] + \left[ 0,1 
ight] = \left[ 1,1 
ight] \notin S^{1}$$

since  $1^2+1^2\neq 1$ 

Since the unit circle is not closed under scalar multiplication and vector addition,  $S_1$  is not a subspace.

Example: the set  $T = \left\{ [x, y] \in \mathbb{R}^2 \mid x, y \in \mathbb{Z} \right\}$ 

The subset T is closed under vector addition: If  $n_1, n_2, m_1, m_2 \in \mathbb{Z}$ , then  $[n_1, n_2]$ ,  $[m_1, m_2] \in T$ . then

$$\mathbf{v}_1 + \mathbf{v}_2 = [n_1 + m_1, n_2 + m_2]$$

Since the sum of two integers is always another integer, both components of the vector sum  $\mathbf{v}_1 + \mathbf{v}_2$  are integers. Thus,  $\mathbf{v}_1 + \mathbf{v}_1$  is always in T; and so, T is closed under vector addition.

Example: the set  $\mathcal{T} = \left\{ [x, y] \in \mathbb{R}^2 \mid x, y \in \mathbb{Z} \right\}$ , Cont'd

However, the set T is not closed under scalar multiplication. To see this, choose  $\lambda = \sqrt{2} \in \mathbb{R}$  and let  $\mathbf{v} = [1,0] \in T$ . Then,

$$\lambda \mathbf{v} = \sqrt{2} \left[ 1, 0 \right] = \left[ \sqrt{2}, 0 \right] \notin T$$
 since  $\sqrt{2}$  is not an integer

Since  $\lambda \mathbf{v} \notin T$ , T is not closed under scalar multiplication.

Since T is not closed under **both** scalar multiplication and vector addition, T is not a subspace.

## Example: A hyperplane in $\mathbb{R}^n$

Consider a hyperplane in  $\mathbb{R}^n$ 

$$\mathcal{H} = \{\mathbf{p}_0 + s_1 \mathbf{w}_1 + \dots + s_k \mathbf{w}_k \mid s_1, \dots, s_k \in \mathbb{R}\}$$

If  $\mathcal{H}$  does not contain the zero vector **0** it is not a subspace. To see this, suppose  $\mathbf{v} \in \mathcal{H}$ , and we choose  $\lambda = 0$ . Then

$$\lambda \mathbf{v} = \mathbf{0}\mathbf{v} = \mathbf{0} \notin \mathcal{H}$$

and so  ${\mathcal H}$  is not closed under scalar multiplication and so  ${\mathcal H}$  is not a subspace.

The preceding counter-example generalizes as follows:

#### Lemma

Let S be a subset of  $\mathbb{R}^n$ . If  $\mathbf{0} \notin S$ , then S is not closed under scalar multiplication and hence S is not a subspace.

(scalar multiplication by 0 would always yield  $\mathbf{0}$  and so would take you out of a set S that doesn't contain  $\mathbf{0}$ )

The Span of a Set of Vectors

Recall

## Definition

Let  $\{\mathbf{w}_1, \ldots, \mathbf{w}_k\}$  be a set of vectors in  $\mathbb{R}^n$ . The **span** of  $\mathbf{w}_1, \ldots, \mathbf{w}_k$  is the set of all linear combinations of  $\mathbf{w}_1, \ldots, \mathbf{w}_k$ :

$$span(\mathbf{w}_1,\ldots,\mathbf{w}_k)=\{s_1\mathbf{w}_1+\cdots+s_k\mathbf{w}_k\mid s_1,\ldots,s_k\in\mathbb{R}\}$$

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# First Basic Prototype of a Subspace: the span of a set of vectors

#### Let

$$S = span(\mathbf{w}_1, \dots, \mathbf{w}_k) \equiv \{s_1\mathbf{w}_1 + \dots + s_k\mathbf{w}_k \mid s_1, \dots, s_k \in \mathbb{R}\}$$

Then if  $\lambda \in \mathbb{R}$  and  $\mathbf{v} = s_1 \mathbf{w}_1 + \cdots + s_k \mathbf{w}_k \in S$ , then

$$\lambda \mathbf{v} = \lambda (s_1 \mathbf{w}_1 + \dots + s_k \mathbf{w}_k) = (\lambda s_1) \mathbf{w}_1 + \dots + (\lambda s_k) \mathbf{w}_k$$

Since each of the scalar factors  $(\lambda s_i)$  on the right is a real number,  $\lambda \mathbf{v}$  is another element of S (for  $\lambda \mathbf{v}$  is just another linear combination of the vectors  $\mathbf{w}_1, \ldots, \mathbf{w}_k$ ).

So the span of a set of vectors is always closed under scalar multiplication.

Example: the span of a set of vectors, Cont'd

$$S = span(\mathbf{w}_1, \dots, \mathbf{w}_k) = \{s_1\mathbf{w}_1 + \dots + s_k\mathbf{w}_k \mid s_1, \dots, s_k \in \mathbb{R}\}$$

Now choose two vectors in S:

$$\mathbf{v}_1 = s_1 \mathbf{w}_1 + \dots + s_k \mathbf{w}_k$$
$$\mathbf{v}_2 = t_1 \mathbf{w}_1 + \dots + t_k \mathbf{w}_k$$

Then

$$\mathbf{v}_1 + \mathbf{v}_2 = (s_1 + t_1) \mathbf{w}_1 + \dots + (s_k + t_k) \mathbf{w}_k$$

Since the scalar factors  $(s_i + t_i)$  on the right are all real numbers,  $\mathbf{v}_1 + \mathbf{v}_2$  belongs to S. Hence,  $S = span(\mathbf{w}_1, \dots, \mathbf{w}_k)$  is closed under vector addition. Example: the span of a set of vectors, Cont'd

 $span(\mathbf{w}_1, \ldots, \mathbf{w}_k)$  is thus closed under both scalar multiplication and vector addition,

If  $\{\mathbf{w}_1, \ldots, \mathbf{w}_k\} \subset \mathbb{R}^n$ , then

 $span(\mathbf{w}_1,\ldots,\mathbf{w}_k) \equiv \{t_1\mathbf{w}_1+\cdots+t_k\mathbf{w}_k \mid t_1,\ldots,t_k \in \mathbb{R}\}$ 

is always a subspace of  $\mathbb{R}^n$ .

# Solution Set of a Linear System

Let S be the solution set of an  $n \times m$  linear system:

$$S = \{\mathbf{y} \in \mathbb{R}^m \mid \mathbf{A}\mathbf{y} = \mathbf{b}\}$$

S is not closed under scalar multiplication: If **y** is a solution, then

$$\mathbf{A}\left(\lambda\mathbf{y}
ight)=\lambda\mathbf{A}\mathbf{y}=\lambda\mathbf{b}
eq\mathbf{b}$$

so  $\lambda \mathbf{y}$  is not a solution

S is not closed under vector addition: If **y** and **w** are solutions, then

$$A(y + w) = Ay + Aw = b + b = 2b \neq b$$

so  $\mathbf{y} + \mathbf{w}$  is not a solution.

Since the solution set is not automatically closed under either scalar multiplication and vector addition,

The solution set of a linear system is a not, in general, a subspace of the vector space of variable values  $\mathbb{R}^m_{-}$ 

# The Solution Set of a Homogeneous Linear System

A homogeneous linear system is a linear system of the form Ax = 0 (where the right hand side is the zero vector).

Let S be the solution set of an  $n \times m$  homogeneous linear system.

S is closed under scalar multiplication: If  $\mathbf{y}$  is a solution, then

$$\mathbf{A}(\lambda \mathbf{y}) = \lambda \mathbf{A} \mathbf{y} = \lambda \mathbf{0} = \mathbf{0}$$

so  $\lambda \mathbf{y}$  is also a solution

S is closed under vector addition: If  $\mathbf{y}$  and  $\mathbf{w}$  are solutions, then

$$A(y+w) = Ay + Aw = 0 + 0 = 0$$

so  $\mathbf{y} + \mathbf{w}$  is a solution.

Since the solution set is closed under both scalar multiplication and vector addition, the solution set of a homogeneous linear system Ax = 0 is a subspace of the vector space of variable values  $\mathbb{R}^m$ .

# Remarks

- ▶ Hyperplanes  $\mathcal{H} = \{\mathbf{p}_0 + s_1 \mathbf{w}_1 + \cdots + s_k \mathbf{w}_k \mid s_1, \ldots, s_k \in \mathbb{R}\}$ are not subspaces in general, but
- Spanning sets  $span(\mathbf{w}_1, \dots, \mathbf{w}_k) = \{s_1\mathbf{w}_1 + \dots + s_k\mathbf{w}_k \mid s_1, \dots, s_k \in \mathbb{R}\}$  are subspaces
- Similarly,
  - Solution sets of linear systems Ax = b are not subspaces in general, but
  - Solution sets of homogeneous linear systems Ax = 0 are subspaces.

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# Summary So Far

#### Recall

## Definition

A subset S of a vector space  $\mathbb{R}^n$  is a **subspace** if:

$$\blacktriangleright \ \mathbf{v} \in S \text{ and } \lambda \in \mathbb{R} \implies \lambda \mathbf{v} \in S$$

 $\blacktriangleright$   $\mathbf{v}_1, \mathbf{v}_2 \in S \implies \mathbf{v}_1 + \mathbf{v}_2 \in S$ 

#### Two basic prototypes for subspaces:

the span of a set of vectors: e.g.,

$$S = span(\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k)$$
  
$$\equiv \{c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_k\mathbf{v}_k \mid c_1, c_2, \dots, c_k \in \mathbb{R}\}$$

• the solution set of a homogeneous linear system Ax = 0We'll next focus on the first prototype.

# Nomenclature

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$$S = span(\mathbf{v}_1, \ldots, \mathbf{v}_k)$$

we say that S is **the subspace generated** by the vectors  $\mathbf{v}_1, \ldots, \mathbf{v}_k$ .

We also say that the vectors  $\mathbf{v}_1, \ldots, \mathbf{v}_k$  are a **set of generators** for *S*.

A key concern for us today is that **the vectors that generate a subspace are far from unique**.

In fact, even the number of generators is not unique for a given subspace.

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# Finding good coordinates for the points of a subspace

Since a subspace is (generally) a lower dimensional hyperplane inside a higher dimensional vector space  $\mathbb{R}^n$ , the usual coordinates for  $\mathbb{R}^n$  are not really suitable coordinates for vectors in a subspace. Let's begin with an example:

## Example 1.

Consider

$$S = span\left( \left[ \begin{array}{c} 1\\1 \end{array} \right], \left[ \begin{array}{c} 1\\-1 \end{array} \right], \left[ \begin{array}{c} 0\\1 \end{array} \right] \right)$$

For each choice of numbers  $c_1, c_2, c_3$ ,

$$\mathbf{v}_{c_1,c_2,c_3} = c_1 \begin{bmatrix} 1\\1 \end{bmatrix} + c_2 \begin{bmatrix} 1\\-1 \end{bmatrix} + c_3 \begin{bmatrix} 0\\1 \end{bmatrix}$$

belongs to S.

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# Example, Cont'd

But the numbers  $c_1$ ,  $c_2$ ,  $c_3$  do not provide good coordinates for the vectors in S since two different choices of  $c_1$ ,  $c_2$ ,  $c_3$  can correspond to the same vector. E.g.,

$$\begin{array}{c} c_{1} = 1 \\ c_{2} = 1 \\ c_{3} = 0 \end{array} \right\} \qquad \Rightarrow \qquad (1) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (1) \begin{bmatrix} 1 \\ -1 \end{bmatrix} + (0) \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$
$$\begin{array}{c} c_{1} = 0 \\ c_{2} = 0 \\ c_{3} = 2 \end{array} \right\} \qquad \Rightarrow \qquad (0) \begin{bmatrix} 1 \\ 1 \end{bmatrix} + (0) \begin{bmatrix} 1 \\ -1 \end{bmatrix} + (2) \begin{bmatrix} 0 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \end{bmatrix}$$

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## Bases

Circumstances like the preceding example motivate the following definition:

## Definition

A **basis** for a subspace S is a set of vectors  $\{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_k\}$  such that **every** vector  $\mathbf{v} \in S$  can be expressed as

$$\mathbf{v} = c_1 \mathbf{b}_1 + c_2 \mathbf{b}_2 + \dots + c_k \mathbf{b}_k$$

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for exactly one choice of coefficients  $c_1, c_2, \ldots, c_k$ .

## Remarks:

If  $\{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_k\}$  is a basis for a subspace S, then

- $\blacktriangleright S = span(\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_k)$
- If two linear combinations of the vectors b<sub>1</sub>, b<sub>2</sub>,..., b<sub>k</sub> yield the same vector, then the coefficients have to be the same:

$$c_1\mathbf{b}_1+\cdots+c_k\mathbf{b}_k=d_1\mathbf{b}_1+\cdots+d_k\mathbf{b}_k$$

requires

$$c_1 = d_1 \ , \ c_2 = d_2 \ , \ \dots \ , c_k = d_k$$

Thus, to each vector in S there corresponds a unique list of numbers [c<sub>1</sub>,..., c<sub>k</sub>]. These numbers provide suitable (i.e., unique) coordinates for each vector of S.

# The Utility of Bases

- If  $B = \{\mathbf{w}_1, \dots, \mathbf{w}\} k\}$  is a basis for a subspace  $S \subset \mathbb{R}^n$ ,
  - Each  $\mathbf{v} \in S$  corresponds to a unique ordered list of k numbers
  - To each element  $\mathbb{R}^k$  corresponds to a unique element of S.

$$\mathbf{v} \in S \implies \mathbf{v} = c_1 \mathbf{w}_1 + \dots + c_k \mathbf{w}_k \iff [c_1, \dots, c_k] \in \mathbb{R}^k$$

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