# MAT224 - LEC5101 - Lecture 2 Linear combinations, span, and independence

Dylan Butson

University of Toronto

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### Subspaces review

**Definition**: A *subspace* of a vector space V is a subset W of V that is itself a vector space with the same operations of vector addition and scalar multiplication as in V.

Note that for  $\mathbf{w}, \mathbf{x} \in W$ , we only know  $\mathbf{w} + \mathbf{x}, c\mathbf{w} \in V$  in general. Thus, the definition of subspace implicitly requires that W satisfies:

- closed under addition: for any  $\mathbf{w}, \mathbf{x} \in W$ ,  $\mathbf{w} + \mathbf{x} \in W$ .
- closed under scalar mult.: for any  $\mathbf{w} \in W$ ,  $c \in \mathbb{R}$ ,  $c\mathbf{w} \in W$ .

Equivalently, we can combine these into a single criterion:

► closed under both: for each 
$$\mathbf{x}, \mathbf{w} \in W$$
 and  $c \in \mathbb{R}$ ,  
 $c\mathbf{w} + \mathbf{x} \in W$ .

**Proposition**: A subset W of V is a subspace if and only if it is non-empty and closed under both addition and scalar mult.

#### Example:

▶ 
$$W_1 = \{(x_1, x_2) \in \mathbb{R}^2 | x_2 = 2x_1\} \subset \mathbb{R}^2 = V$$
 is a subspace.  
▶  $W_2 = \{(x_1, x_2) \in \mathbb{R}^2 | x_2 = x_1^2\} \subset \mathbb{R}^2 = V$  is not.

Discussion: Properties and (non) examples of subspaces

Let  $A \in M_{m \times n}(\mathbb{R})$ . Show that

- (1) the null space  $\operatorname{null}(A) = \{ \mathbf{x} \in \mathbb{R}^n \mid A\mathbf{x} = \mathbf{0} \}$  is a subspace of  $\mathbb{R}^n$ ; and
- (2) the column space  $\operatorname{col}(A) = \{A\mathbf{x} \in \mathbb{R}^m \mid \mathbf{x} \in \mathbb{R}^n\}$  is a subspace of  $\mathbb{R}^m$ .

Which of the following subsets W of  $M_{n \times n}(\mathbb{R})$  are subspaces of  $M_{n \times n}(\mathbb{R})$ ?

(3) 
$$W = \{A \in M_{n \times n}(\mathbb{R}) \mid A \text{ is invertible}\}$$
  
(4)  $W = \{A \in M_{n \times n}(\mathbb{R}) \mid \text{ the last column of } A \text{ is zero}\}$   
(5)  $W = \{A \in M_{n \times n}(\mathbb{R}) \mid A^2 = \mathbf{0}\}$ 

(6) Give examples of subsets W ⊂ ℝ<sup>2</sup> that are closed under addition but not scalar multiplication, and vice versa.

### Linear Combination and Span

**Definition:** Let  $S = {x_1, ..., x_n}$  be a (finite) subset of V.

- ► A vector  $\mathbf{v} \in V$  is a *linear combination* of vectors in S if  $\mathbf{v} = a_1 \mathbf{x}_1 + ... + a_n \mathbf{x}_n$  for some  $a_1, ..., a_n \in \mathbb{R}$ .
- Span(S) is the set of all linear combinations of vectors in S: Span(S) = { $\mathbf{v} = a_1\mathbf{x}_1 + ... + a_n\mathbf{x}_n \in V | a_1, ..., a_n \in \mathbb{R}$ }  $\subset V$
- If Span(S) = W we say S spans the subspace W.

#### Examples:

Let  $\mathbf{x} = (1,0) \in \mathbb{R}^2$ . Then Span $(\mathbf{x}) = \{\mathbf{v} = (v_1,0) \in \mathbb{R}^2\} = \{\mathbf{v} = (v_1,v_2) \in \mathbb{R}^2 | v_2 = 0\}$ Let  $\mathbf{x}_1 = (1,2) \in \mathbb{R}^2$ . Then Span $(\mathbf{x}_1) = \{\mathbf{v} = (v_1, 2v_1) \in \mathbb{R}^2\} = \{\mathbf{v} = (v_1, v_2) \in \mathbb{R}^2 | v_2 = 2v_1\}$ Let  $\mathbf{x}_2 = (2,4) \in \mathbb{R}^2$ . Then Span $(\mathbf{x}_2) = \text{Span}(\mathbf{x}_1) = \text{Span}(\{\mathbf{x}_1, \mathbf{x}_2\})$ .

#### **Proposition:**

(a) The set  $\text{Span}(S) \subset V$  defines a subspace of V.

**(b)** If  $W \subset V$  is a subspace of V and  $S \subset W$ , then  $\text{Span}(S) \subset W$ .

Discussion: Examples of Spans and the Sum Operation

Let 
$$\mathbf{v} = (1, 0, 0), \mathbf{w} = (0, 1, 2), \mathbf{x} = (1, 1, 2) \in \mathbb{R}^3.$$

(1) Let  $S = {\mathbf{v}, \mathbf{w}}$ , describe Span(S) as a subset of  $\mathbb{R}^3$  as above.

(2) Let 
$$S = {\mathbf{v}, \mathbf{x}}$$
, and do the same.

(3) How do these compare with Span({v, w, x})?

Let 
$$S = \{1, x - 2x^2, 3x^2, 1 + 4x^2\} \subset P_2(\mathbb{R})$$
  
(4) Show  $\{1, x, x^2\} \subseteq \text{Span}(S) \subseteq P_2(\mathbb{R})$ .  
(5) Deduce that  $P_2(\mathbb{R}) \subseteq \text{Span}(S)$  and thus  $\text{Span}(S) = P_2(\mathbb{R})$ .

**Bonus:** Let W and X be subspaces of V and define  $W + X = \{\mathbf{v} = \mathbf{w} + \mathbf{x} \in V | \mathbf{w} \in W \text{ and } \mathbf{x} \in X\} \subset V$ 

(a) Show that W + X is a subspace of V.
(b) If W = Span({w<sub>1</sub>, ..., w<sub>n</sub>}) and X = Span({x<sub>1</sub>, ..., x<sub>m</sub>}) then W + X = Span({w<sub>1</sub>, ..., w<sub>n</sub>, x<sub>1</sub>, ..., x<sub>m</sub>})

### Linear (In)Dependence

**Definition:** Let  $S = {x_1, ..., x_n}$  be a subset of V.

▶ A *linear dependence* among the vectors of S is an equation:

 $a_1\mathbf{x}_1 + ... + a_n\mathbf{x}_n = \mathbf{0}$  for some  $a_1, ..., a_n \in \mathbb{R}$ 

It is called *trivial* if  $a_1 = ... = a_n = 0$  and *non trivial* otherwise.

S is called *linearly dependent* if it has a non trivial linear dependence.

**Example:** Let  $\mathbf{v} = (1, 0, 0), \mathbf{w} = (0, 1, 2), \mathbf{x} = (1, 1, 2) \in \mathbb{R}^3$ .

• The set  $S = \{v, w, x\}$  has a non trivial linear dependence:

$$1\mathbf{v} + 1\mathbf{w} + (-1)\mathbf{x} = \mathbf{0}$$

**Definition:** *S* is called *linearly independent* if it has no non trivial linear dependence.

Equivalently, S called is linearly independent if: for any equation

$$a_1\mathbf{x}_1 + ... + a_n\mathbf{x}_n = \mathbf{0}$$
 for some  $a_1, ..., a_n \in \mathbb{R}$ 

we must have  $a_1 = \ldots = a_n = 0$ .

**Example:** The set  $S_1 = \{\mathbf{v}, \mathbf{w}\}$  is linearly independent.

## Discussion: Examples of Linear (In)dependence

(1) Let 
$$S = \{1 + x, x + x^2, 1 - x^2\} \subset P_2(\mathbb{R})$$
.  
Prove that S is linearly dependent.

(2) Let 
$$S = {\mathbf{x}_1, ..., \mathbf{x}_n} \subset V$$
 and  $\mathbf{v} \in \text{Span}(S)$ .  
Prove that  $S' = {\mathbf{x}_1, ..., \mathbf{x}_n, \mathbf{v}}$  is linearly dependent.

(3) Let 
$$e_1 = (1, 0, 0), e_2 = (0, 1, 0), e_3 = (0, 0, 1) \in \mathbb{R}^3$$
.  
Prove that  $S = \{e_1, e_2, e_3\}$  is linearly independent.

(4) Let 
$$S = \{1 + x, x + x^2, 1 + x^2\} \subset P_2(\mathbb{R})$$
.  
Prove that S is linearly independent.

**Bonus:** Let  $S = \{\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3\}$  and  $S_1 = \{\mathbf{x}_2, \mathbf{x}_3\}, S_2 = \{\mathbf{x}_1, \mathbf{x}_3\}$ , and  $S_3 = \{\mathbf{x}_1, \mathbf{x}_2\}$ . If  $S_1, S_2, S_3$  are each linearly independent, does this imply S is linearly independent? Prove or give counterexample.