Linear algebra: Core definitions

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#### **Notation**

Linear algebra: Core definitions

- $\mathbb{R}^m$  and  $\mathbb{C}^m$  are the m-dimensional spaces of real and complex numbers, respectively.
  - The term "scalar" will refer to a member of  $\mathbb R$  or  $\mathbb C$  depending whether we're working in  $\mathbb R^m$  or  $\mathbb C^m$ .
- In what follows,  $S=\left\{\mathbf{a}_j\right\}_{j=1}^n$  is a set of vectors, with each  $\mathbf{a}_j\in\mathbb{C}^m$  or  $\mathbb{R}^m.$

## Definition: Linear combination and span

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#### Definition (Linear combination and span)

A *linear combination* (LC) of the members of S is a scalar-weighted sum such as

$$\mathbf{a}_1x_1+\mathbf{a}_2x_2+\cdots+\mathbf{a}_nx_n,$$

where the n coefficients  $\{x_j\}_{j=1}^m$  are scalars. The **span** of S, denoted span S, is the set of all linear combinations that can be formed by the members of S.

# Definition: Vector space and subspace

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#### Definition (Vector space and subspace)

A set V of vectors is called a **vector space** if it obeys the axioms listed in Appendix B of the notes. We needn't go into all the axioms for this course; as long as one defines addition and scalar multiplication "reasonably," then it is sufficient to consider the **closure conditions** 

- $\mathbf{x} + \mathbf{y} \in V$  for all  $\mathbf{x}, \mathbf{y} \in V$
- $\bullet \ \alpha \mathbf{x} \in V \ \text{for all} \ \mathbf{x} \in V \ \text{and for all scalars} \ \alpha$

and the zero condition,

V contains the zero vector 0.

A subset of V that is itself a vector space is called a *subspace*.

## Definition: Matrix-vector product

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#### Definition

Form the matrix  $A \in \mathbb{C}^{m \times n}$  by packing the members of S into the columns of A. Given a vector  $\mathbf{x} \in \mathbb{C}^n$ , the **matrix-vector product**  $A\mathbf{x}$  is the linear combination of the  $\mathbf{a}$ 's weighted by the entries in  $\mathbf{x}$ ,

$$A\mathbf{x} = \mathbf{a}_1 x_1 + \mathbf{a}_2 x_2 + \cdots + \mathbf{a}_n x_n.$$

#### Example: Matrix-vector product

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$$ullet$$
 Let  $A=\left[egin{array}{cc} -2 & 1 \ 3 & 5 \end{array}
ight]$  and  ${f x}=\left[egin{array}{cc} 4 & 7 \end{array}
ight]^T$ . Then

$$A\mathbf{x} = 4 \begin{bmatrix} -2 \\ 3 \end{bmatrix} + 7 \begin{bmatrix} 1 \\ 5 \end{bmatrix} = \begin{bmatrix} -1 \\ 47 \end{bmatrix}.$$

Compare to the "dot product" approach,

$$A\mathbf{x} = \begin{bmatrix} -2 \cdot 4 + 1 \cdot 7 \\ 3 \cdot 4 + 5 \cdot 7 \end{bmatrix} = \begin{bmatrix} -1 \\ 47 \end{bmatrix}.$$

# Definition: Linear dependence and independence

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#### Definition

The set S is called *linearly dependent* (LD) if there is a *nonzero* vector  $\mathbf{x} \in \mathbb{C}^n$  such that

$$\mathbf{a}_1x_1+\mathbf{a}_2x_2+\cdots+\mathbf{a}_nx_n=\mathbf{0}.$$

If no such vector exists, then we call S linearly independent (LI).

### Example: Linear dependence and independence

Linear algebra: Core definitions

- Let  $\mathbf{a}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $\mathbf{a}_2 = \begin{bmatrix} 2 \\ 4 \end{bmatrix}$ . The set  $\{\mathbf{a}_1, \mathbf{a}_2\}$  is LD because  $\mathbf{a}_1 x_1 + \mathbf{a}_2 x_2 = \mathbf{0}$  for nonzero vector  $\mathbf{x} = \begin{bmatrix} -2 \\ 1 \end{bmatrix} s$ , where s is any constant.
- Let  $\mathbf{a}_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $\mathbf{a}_2 = \begin{bmatrix} 2 \\ 2 \end{bmatrix}$ . The set  $\{\mathbf{a}_1, \mathbf{a}_2\}$  is LI because the linear combination

$$\mathbf{a}_1 x_1 + \mathbf{a}_2 x_2 = \left[ \begin{array}{c} x_1 + 2x_2 \\ 2x_1 + 2x_2 \end{array} \right]$$

cannot be zero unless  $x_1 = x_2 = 0$ .

### Definition: Null space of a matrix

Linear algebra: Core definitions

#### Definition (Null space)

The *null space* of a matrix  $A \in \mathbb{C}^{m \times n}$ , denoted null (A), is the set of vectors  $\mathbf{x} \in \mathbb{C}^n$  such that  $A\mathbf{x} = \mathbf{0}$ .

- Notice that null (A) always contains the zero vector. If null (A) contains only the zero vector, that null space is called trivial; otherwise, it is called nontrivial.
- It's not hard to show that null(A) obeys the closure conditions and is therefore a vector space.
- Notice that null (A) is trivial iff the columns of A are LI.

### Example: Null spaces

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• Let  $A=\begin{bmatrix}1&2\\2&4\end{bmatrix}$ . Then  $\operatorname{null}(A)$  consists of all scalar multiples of  $\begin{bmatrix}-2\\1\end{bmatrix}$ ,

$$\operatorname{null}(A) = \left\{ c \begin{bmatrix} -2 \\ 1 \end{bmatrix} : c \in \mathbb{R} \right\}.$$

• Let  $A = \begin{bmatrix} 1 & 2 \\ 2 & 2 \end{bmatrix}$ . Then null (A) is trivial.

# Definition: Range space of a matrix

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#### Definition (Range space)

The **range space** of a matrix  $A \in \mathbb{C}^{m \times n}$ , denoted range (A), is the set of vectors  $\mathbf{y} \in \mathbb{C}^m$  that can be formed by linear combination of the columns of A. In formal notation,

range 
$$(A) = \{ \mathbf{y} \in \mathbb{C}^m : \exists \mathbf{x} \in \mathbb{C}^n \text{ for which } A\mathbf{x} = \mathbf{y} \}.$$

 It's not hard to show that range (A) obeys the closure conditions and is therefore a vector space. • Let  $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$ . Then range (A) consists of all scalar multiples of  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,

range 
$$(A) = \left\{ c \begin{bmatrix} 1 \\ 2 \end{bmatrix} : c \in \mathbb{R} \right\}.$$

• Let  $A = \begin{bmatrix} 1 & 2 \\ 2 & 2 \end{bmatrix}$ . Then range (A) consists of all linear combinations  $x_1 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + x_2 \begin{bmatrix} 2 \\ 2 \end{bmatrix}$ , which is  $\mathbb{R}^2$ .