

- New concepts:
  - Span of a vector set, matrix column space (range)
  - Linearly dependent set of vectors
  - Matrix null space

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 1 & 2 & 3 \end{pmatrix} \in \mathbb{R}^{3 \times 3}, \mathbf{b} = \begin{pmatrix} 3 \\ 1 \\ 3 \end{pmatrix} \in \mathbb{R}^3, \mathbf{c} = \begin{pmatrix} 3 \\ 1 \\ 4 \end{pmatrix} \in \mathbb{R}^3 \quad (1)$$

$$\begin{cases} x_1 + 2x_2 + 3x_3 = 3 \\ x_2 + x_3 = 1 \\ x_1 + 2x_2 + 3x_3 = 3 \end{cases} \Leftrightarrow \mathbf{Ax} = \mathbf{b}, \begin{cases} y_1 + 2y_2 + 3y_3 = 3 \\ y_2 + y_3 = 1 \\ y_1 + 2y_2 + 3y_3 = 4 \end{cases} \Leftrightarrow \mathbf{Ay} = \mathbf{c}$$

Form the bordered matrix in both cases, and reduce to triangular form (Gaussian elimination)

$$\left( \begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 1 & 2 & 3 & 3 \end{array} \right) \sim \left( \begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 \end{array} \right) \Leftrightarrow \begin{cases} x_1 + 2x_2 + 3x_3 = 3 \\ x_2 + x_3 = 1 \\ 0 = 0 \end{cases} \text{ Infinite number of solutions}$$

$$\left( \begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 1 & 2 & 3 & 4 \end{array} \right) \sim \left( \begin{array}{ccc|c} 1 & 2 & 3 & 3 \\ 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 1 \end{array} \right) \Leftrightarrow \begin{cases} y_1 + 2y_2 + 3y_3 = 3 \\ y_2 = 1 \\ 0 \neq 1 \end{cases} \text{ No solutions}$$

Recall: Solving  $\mathbf{A}\mathbf{x} = \mathbf{b}$  means finding the linear combination of columns of  $\mathbf{A}$  such that

$$x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + x_3\mathbf{a}_3 = \mathbf{b}, \mathbf{x} = (x_1 \ x_2 \ x_3)^T \quad (2)$$

For previous examples

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 1 & 2 & 3 \end{pmatrix} \Rightarrow \mathbf{a}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \mathbf{a}_2 = \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}, \mathbf{a}_3 = \begin{pmatrix} 3 \\ 1 \\ 3 \end{pmatrix}$$

Note that  $\mathbf{a}_3 = \mathbf{a}_1 + \mathbf{a}_2$ , so the linear combination (2) can be rewritten as

$$(x_1 + x_3)\mathbf{a}_1 + (x_2 + x_3)\mathbf{a}_2 = \mathbf{b}$$

If  $\mathbf{b}$  is in the plane defined by the two directions  $\mathbf{a}_1, \mathbf{a}_2$  then there are an infinity of choices of  $\mathbf{x} = (x_1 \ x_2 \ x_3)^T$  to obtain  $\mathbf{b}$  by the linear combination of  $\mathbf{a}_1, \mathbf{a}_2$ . This is the first example.

Since  $\mathbf{c} = (3 \ 1 \ 4)^T$  is not in the plane defined by  $\mathbf{a}_1, \mathbf{a}_2$ , there is no linear combination of  $\mathbf{a}_1, \mathbf{a}_2$  to obtain  $\mathbf{c}$ , hence there is no solution to the system  $\mathbf{A}\mathbf{y} = \mathbf{c}$ .

- Generalize this idea of “vectors reachable by linear combination of other vectors”

**Definition.** The *span* of vectors  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n \in \mathcal{V}$ , is the set of vectors reachable by linear combination

$$\text{span}\{\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n\} = \{\mathbf{b} \in \mathcal{V} \mid \exists x_1, \dots, x_n \in \mathcal{S} \text{ such that } \mathbf{b} = x_1\mathbf{a}_1 + \dots x_n\mathbf{a}_n\}.$$

The notation used for set on the right hand side is read: “those vectors  $\mathbf{b}$  in  $\mathcal{V}$  with the property that there exist  $n$  scalars  $x_1, \dots, x_n$  to obtain  $\mathbf{b}$  by linear combination of  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$ .”

- A linear combination is conveniently expressed as a matrix-vector product leading to a different formulation of the same concept

**Definition.** The *column space* (or *range*) of matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  is the set of vectors reachable by linear combination of the matrix column vectors

$$C(\mathbf{A}) = \text{range}(\mathbf{A}) = \{\mathbf{b} \in \mathbb{R}^m \mid \exists \mathbf{x} \in \mathbb{R}^n \text{ such that } \mathbf{b} = \mathbf{A}\mathbf{x}\} \subseteq \mathbb{R}^m$$

In the example (1)

$$\mathbf{A} = (\mathbf{a}_1 \ \mathbf{a}_2 \ \mathbf{a}_3) = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 1 & 2 & 3 \end{pmatrix}$$

$$\text{span}\{\mathbf{a}_1, \mathbf{a}_2, \mathbf{a}_3\} = \text{span}\{\mathbf{a}_1, \mathbf{a}_2\}$$

since  $\mathbf{a}_3 = \mathbf{a}_1 + \mathbf{a}_2 \Leftrightarrow \mathbf{a}_1 + \mathbf{a}_2 - \mathbf{a}_3 = \mathbf{0}$ . Introduce a concept to capture the idea that a vector can be expressed in terms of other vectors.

**Definition.** The vectors  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n \in \mathcal{V}$ , are *linearly dependent* if there exist  $n$  scalars,  $x_1, \dots, x_n \in \mathcal{S}$ , at least one of which is different from zero such that

$$x_1\mathbf{a}_1 + \dots x_n\mathbf{a}_n = \mathbf{0}$$

Note that  $\{\mathbf{0}\}$ , with  $\mathbf{0} \in \mathcal{V}$  is a linearly dependent set of vectors since  $1 \cdot \mathbf{0} = \mathbf{0}$ .

The converse of linear dependence is linear independence, a member of the set cannot be expressed as a non-trivial linear combination of the other vectors

**Definition.** The vectors  $\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n \in \mathcal{V}$ , are *linearly independent* if the *only*  $n$  scalars,  $x_1, \dots, x_n \in \mathcal{S}$ , that satisfy

$$x_1 \mathbf{a}_1 + \dots x_n \mathbf{a}_n = \mathbf{0}, \quad (3)$$

are  $x_1 = 0, x_2 = 0, \dots, x_n = 0$ .

The choice  $\mathbf{x} = (x_1 \dots x_n)^T = \mathbf{0}$  that always satisfies (3) is called a *trivial solution*. We can restate linear independence as (3) being satisfied *only* by the trivial solution.

Introduce a characterization of the column vectors of a matrix related to linear dependence

**Definition.** The *null space* of a matrix  $\mathbf{A} \in \mathbb{R}^{m \times n}$  is the set

$$N(\mathbf{A}) = \text{null}(\mathbf{A}) = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{A}\mathbf{x} = \mathbf{0}\} \subseteq \mathbb{R}^n$$

- If  $\text{null}(\mathbf{A}) = \{\mathbf{0}\}$  then the column vectors of  $\mathbf{A}$  are linearly independent, since the only way to satisfy (3) is by the trivial solution  $\mathbf{x} = \mathbf{0}$

For example (1) we have  $c(\mathbf{a}_1 + \mathbf{a}_2 - \mathbf{a}_3) = \mathbf{0}$  for any scalar  $c$ , hence

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 1 & 2 & 3 \end{pmatrix} \Rightarrow \mathbf{a}_1 = \begin{pmatrix} 1 \\ 0 \\ 1 \end{pmatrix}, \mathbf{a}_2 = \begin{pmatrix} 2 \\ 1 \\ 2 \end{pmatrix}, \mathbf{a}_3 = \begin{pmatrix} 3 \\ 1 \\ 3 \end{pmatrix}$$

$$C(\mathbf{A}) = \text{span}\{\mathbf{a}_1, \mathbf{a}_2\}, N(\mathbf{A}) = \text{span}\left\{\begin{pmatrix} 1 \\ 1 \\ -1 \end{pmatrix}\right\}$$