

# SC223 - Linear Algebra

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Lecture 4



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## When does a solution exist?

- ▶ Solution to  $Ax = b$  exists if and only if  $b$  belongs to the set of *all possible linear combinations* of columns of  $A$ .
- ▶ **Column Space:** The set of all possible linear combinations of columns of  $A$  is called the Column space of matrix  $A$ , and is denoted by  $C(A)$ .

$$C(A) = \{Ax \mid \forall x \in \mathbb{R}^n\}, C(A) \subseteq \mathbb{R}^m$$

- ▶ *Properties:*

- ▶ Let  $\mathbf{0}_m = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}_{m \times 1}$ .  $\mathbf{0}_m \in C(A)$  for any matrix  $A$

- ▶ If  $b_1, b_2 \in C(A)$ ,  $\forall p, q \in \mathbb{R}$ ,  $p \cdot b_1 + q \cdot b_2 \in C(A)$ .

## What about multiple solutions?

- ▶ Let  $Ax = b$  and  $Ay = b$ , with  $x \neq y$ .
- ▶ Then,  $A(x - y) = \mathbf{0}_m$ .
- ▶ Similarly, let  $z \in \mathbb{R}^n$  be such that  $Az = \mathbf{0}_m$ . Then, if  $Ax = b$ ,  $A(x + z) = b \Rightarrow$  Multiple Solutions!
- ▶ **Nullspace:** For a matrix  $A \in \mathbb{R}^{m \times n}$ , the *Nullspace* is the set of vectors that get mapped to  $\mathbf{0}_m$ , and is denoted by  $N(A)$ .

$$N(A) := \{x \in \mathbb{R}^n \mid Ax = \mathbf{0}_m\}, N(A) \subseteq \mathbb{R}^n$$

- ▶  $Ax = \mathbf{0}_m$  are also called **Homogeneous equations**.
- ▶ *Properties:*
- ▶  $\mathbf{0}_n \in N(A)$ .
- ▶ If  $x, y \in N(A)$ ,  $\forall p, q \in \mathbb{R}, p \cdot x + q \cdot y \in N(A)$ .
- ▶ If  $\exists z \in N(A), z \neq \mathbf{0}_n$ , then  $Ax = b$  will have infinitely many solutions, if one exists!

- If there are multiple solutions to  $Ax = b \Rightarrow \exists z \neq \mathbf{0}_n$  such that  $Az = \mathbf{0}_m$ .
- Re-writing, we get  $\sum_{i=1}^n z_i a_{*i} = \mathbf{0}_m$ , with at least one non-zero entry in  $z$ , say  $z_k$ .

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- **Matrix Transpose:** For  $A \in \mathbb{R}^{m \times n}$  given by

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- **Beware of the notation:**  $a_{i*}$  denotes the  $i^{th}$  row of  $A$  written as a column matrix.

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$$\begin{bmatrix} a_{1*}^T \\ a_{2*}^T \\ \vdots \\ a_{m*}^T \end{bmatrix} = \begin{bmatrix} a_{*1} & \dots & a_{*r} \end{bmatrix}_{m \times r} \begin{bmatrix} c_{1*}^T \\ c_{2*}^T \\ \vdots \\ c_{r*}^T \end{bmatrix}_{r \times n}$$