# Lecture 10: Orthogonality (Section 3.1)

Thang Huynh, UC San Diego 1/31/2018

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reflects every vector in  $\mathbb{R}^2$  through the line y = x.

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$$A = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$
 gives the map  $\begin{bmatrix} x \\ y \end{bmatrix} \mapsto \begin{bmatrix} -y \\ x \end{bmatrix}$ , i.e. rotates every vector in  $\mathbb{R}^2$  counter-clockwise by  $90^\circ$ .

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▶ Example. Find the inner product of the following two vectors

$$\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix}.$$

▶ Definition. The **norm** or **length** of a vector v in  $\mathbb{R}^n$  is

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▶ Example. In  $\mathbb{R}^2$ ,

dist 
$$\begin{bmatrix} x_1 \\ y_1 \end{bmatrix}$$
,  $\begin{bmatrix} x_2 \\ y_2 \end{bmatrix}$  =  $\sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}$ .

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$$\bullet \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} -2 \\ 1 \\ 1 \end{bmatrix}$$

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Theorem. Suppose that  $v_1, ..., v_n$  are nonzero and pairwise orthogonal. Then  $v_1, ..., v_n$  are linearly independent.

▶ Proof.

**Example**. The vectors  $\begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$  are pairwise orthogonal and have length 1.

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$$A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \\ 3 & 6 \end{bmatrix}$$
. Find  $N(A)$  and  $C(A^T)$ .

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- ► Example. Consider  $A = \begin{bmatrix} 1 & 2 \\ 2 & 4 \\ 3 & 6 \end{bmatrix}$ . Find N(A) and  $C(A^T)$ .
- ► Solution.  $N(A) = \operatorname{span}\left\{\begin{bmatrix} -2\\1 \end{bmatrix}\right\}$  and  $C(A^T) = \operatorname{span}\left\{\begin{bmatrix} 1\\2 \end{bmatrix}\right\}$ .

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that the two basis vectors are orthogonal  $\begin{bmatrix} -2 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 2 \end{bmatrix} = 0$ .

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- **Example.** Consider  $A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 0 \\ 3 & 6 & 0 \end{bmatrix}$ . Find N(A) and  $C(A^T)$ .
- Solution.  $N(A) = \operatorname{span} \left\{ \begin{bmatrix} -2\\1\\0 \end{bmatrix} \right\}$  and  $C(A^T) = \operatorname{span} \left\{ \begin{bmatrix} 1\\2\\0 \end{bmatrix}, \begin{bmatrix} 0\\0\\1 \end{bmatrix} \right\}$ .

- Example. Consider  $A = \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 0 \\ 3 & 6 & 0 \end{bmatrix}$ . Find N(A) and  $C(A^T)$ .

the vector in N(A) is orthogonal to the vectors in  $C(A^T)$ 

$$\begin{bmatrix} -2\\1\\0 \end{bmatrix} \cdot \begin{bmatrix} 1\\2\\0 \end{bmatrix} = 0, \quad \begin{bmatrix} -2\\1\\0 \end{bmatrix} \cdot \begin{bmatrix} 0\\0\\1 \end{bmatrix} = 0.$$

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Vectors in N(A) are orthogonal to vectors in  $C(A^T)$ .

- ▶ Definition. Let *W* be a subspace of  $\mathbb{R}^n$ , and  $\nu$  in  $\mathbb{R}^n$ .
  - v is **orthogonal** to W, if  $v \cdot w = 0$  for all w in W.

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  - The **orthogonal complement** of W is the space  $W^{\perp}$  of all vectors in  $\mathbb{R}^n$  that are orthogonal to W.

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- ► Example. In previous example,  $N(A) = \text{span} \left\{ \begin{bmatrix} -2\\1\\0 \end{bmatrix} \right\}$  and

$$C(A^T) = \operatorname{span} \left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \right\}$$
 are orthogonal subspaces.

Theorem. (Fundamental Theorem of Linear Algebra, Part II)

- N(A) is orthogonal to  $C(A^T)$ . (The two spaces are orthogonal complements)
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▶ Proof.

Theorem. (Fundamental Theorem of Linear Algebra, Part II)

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#### ▶ Proof.

Theorem. (Fundamental Theorem of Linear Algebra, Part I) Let A be an  $m \times n$  matrix of rank r.

- $\dim C(A) = \dim C(A^T) = r$ .
- $\dim N(A) = n r$
- $\dim N(A^T) = m r$ .

**Example.** Find all vectors orthogonal to  $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$  and  $\begin{bmatrix} 0 \\ 1 \\ 1 \end{bmatrix}$ .

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- ► Example. Let  $V = \left\{ \begin{bmatrix} a \\ b \\ c \end{bmatrix} : a + b = 2c \right\}$ . Find a basis for the orthogonal complement of V.

#### A new perspective on Ax = b

Ax = b is solvable

 $\iff$  **b** is in C(A)

 $\iff$  **b** is orthogonal to  $N(A^T)$ 

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$$A\mathbf{x} = \mathbf{b}$$
 is solvable  
 $\iff \mathbf{b}$  is in  $C(A)$   
 $\iff \mathbf{b}$  is orthogonal to  $N(A^T)$ 

Example. Let 
$$A = \begin{bmatrix} 1 & 2 \\ 3 & 1 \\ 0 & 5 \end{bmatrix}$$
. For which  $\boldsymbol{b}$  does  $A\boldsymbol{x} = \boldsymbol{b}$  have a solution?

solution?