

## **Fundamental Theorem of Linear Algebra**

- Orthogonal Vectors
- Orthogonal and Orthonormal Set
- Orthogonal Complement of a Subspace  $W$
- Column Space, Row Space and Null Space of a Matrix  $A$
- The **Fundamental Theorem of Linear Algebra**

## Orthogonality

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### Definition 1 (Orthogonal Vectors)

Two vectors  $\vec{u}$ ,  $\vec{v}$  are said to be **orthogonal** provided their dot product is zero:

$$\vec{u} \cdot \vec{v} = 0.$$

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If both vectors are nonzero (not required in the definition), then the angle  $\theta$  between the two vectors is determined by

$$\cos \theta = \frac{\vec{u} \cdot \vec{v}}{\|\vec{u}\| \|\vec{v}\|} = 0,$$

which implies  $\theta = 90^\circ$ . In short, orthogonal vectors form a right angle.

## Orthogonal and Orthonormal Set

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### Definition 2 (Orthogonal Set of Vectors)

A given set of nonzero vectors  $\vec{u}_1, \dots, \vec{u}_k$  that satisfies the **orthogonality condition**

$$\vec{u}_i \cdot \vec{u}_j = 0, \quad i \neq j,$$

is called an **orthogonal set**.

### Definition 3 (Orthonormal Set of Vectors)

A given set of **unit vectors**  $\vec{u}_1, \dots, \vec{u}_k$  that satisfies the **orthogonality condition** is called an **orthonormal set**.

## Orthogonal Complement $W^\perp$ of a Subspace $W$

**Definition.** Let  $W$  be a subspace of an inner product space  $V$ , inner product  $\langle \vec{u}, \vec{v} \rangle$ . The **orthogonal complement** of  $W$ , denoted  $W^\perp$ , is the set of all vectors  $\vec{v}$  in  $V$  such that  $\langle \vec{u}, \vec{v} \rangle = 0$  for all  $\vec{u}$  in  $W$ . In set notation:

$$W^\perp = \{ \vec{v} : \langle \vec{u}, \vec{v} \rangle = 0 \text{ for all } \vec{u} \text{ in } W \}$$

**Example.** If  $V = R^3$  and  $W = \text{span}\{\vec{u}_1, \vec{u}_2\}$ , then  $W^\perp$  is the span of the calculus/physics cross product  $\vec{u}_1 \times \vec{u}_2$ . The equation  $\dim(W) + \dim(W^\perp) = 3$  holds (in general  $\dim(W) + \dim(W^\perp) = \dim(V)$ ).

**Theorem.** If  $W$  is the span of the columns  $\vec{u}_1, \dots, \vec{u}_n$  of  $m \times n$  matrix  $A$  (the column space of  $A$ ), then

$$W^\perp = \text{nullspace}(A^T) = \text{span}\{\text{Strang's Special Solutions for } A^T \vec{u} = \vec{0}\}.$$

**Proof.** Given  $W = \text{span}\{\vec{u}_1, \dots, \vec{u}_n\}$ , then

$$\begin{aligned} W^\perp &= \{ \vec{v} : \vec{v} \cdot \vec{w} = 0, \quad \text{all } \vec{w} \in W \} \\ &= \{ \vec{v} : \vec{u}_j \cdot \vec{v} = 0, \quad j = 1, \dots, n \} \\ &= \{ \vec{v} : A^T \vec{v} = \vec{0} \}. \end{aligned}$$

Strang's Special solutions are a basis for the homogeneous problem  $A^T \vec{u} = \vec{0}$ . Therefore,  $W^\perp = \text{nullspace}(A^T) = \text{span}\{\text{Strang's Special Solutions for } A^T \vec{u} = \vec{0}\}.$

## Column Space, Row Space and Null Space of a Matrix $A$

The column space, row space and null space of an  $m \times n$  matrix  $A$  are sets in  $\mathcal{R}^n$  or  $\mathcal{R}^m$ , defined to be the **span** of a certain set of vectors. The **span theorem** implies that each of these three sets are subspaces.

**Definition.** The **Column Space** of a matrix  $A$  is the span of the columns of  $A$ , a subspace of  $\mathcal{R}^m$ . The **Pivot Theorem** implies that

$$\text{colspace}(A) = \text{span}\{\text{pivot columns of } A\}.$$

**Definition.** The **Row Space** of a matrix  $A$  is the span of the rows of  $A$ , a subspace of  $\mathcal{R}^n$ . The definition implies two possible bases for this subspace, just one selected in an application:

$$\text{rowspace}(A) = \text{span}\{\text{Nonzero rows of rref}(A)\} = \text{span}\{\text{pivot columns of } A^T\}.$$

**Definition.** The **Null Space** of a matrix  $A$  is the set of all solutions  $\vec{x}$  to the homogeneous problem  $A\vec{x} = \vec{0}$ , a subspace of  $\mathcal{R}^n$ . Because solution  $\vec{x}$  of  $A\vec{x} = \vec{0}$  is a linear combination of Strang's special solutions, then

$$\text{nullspace}(A) = \text{span}\{\text{Strang's Special Solutions for } A\vec{x} = \vec{0}\}.$$

## The Row space is orthogonal to the Null Space

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**Theorem.** Each row vector  $\vec{r}$  in matrix  $A$  satisfies  $\vec{r} \cdot \vec{x} = 0$ , where  $\vec{x}$  is a solution of the homogeneous equation  $A\vec{x} = \vec{0}$ . Therefore

$$\text{rowspace}(A) \perp \text{nullspace}(A).$$

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The theorem is remembered from this diagram:

$$\begin{pmatrix} \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \\ \cdot & \cdot & \cdot \end{pmatrix} \vec{x} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} \quad \text{is equivalent to} \quad \begin{pmatrix} \text{row 1} \cdot \vec{x} \\ \text{row 2} \cdot \vec{x} \\ \text{row 3} \cdot \vec{x} \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

which says that the rows of  $A$  are orthogonal to solutions  $\vec{x}$  of  $A\vec{x} = \vec{0}$ .

## Computing the Orthogonal Complement of a subspace $W$

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**Theorem.** In case  $W$  is the subspace of  $\mathbf{R}^3$  spanned by two independent vectors  $\vec{u}_1, \vec{u}_2$ , then the orthogonal complement of  $W$  is the line through the origin generated by the cross product vector  $\vec{u}_1 \times \vec{u}_2$ :

$$W^\perp = \text{span}\{\vec{u}_1, \vec{u}_2\}^\perp = \text{span}\{\vec{u}_1 \times \vec{u}_2\}$$

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**Theorem.** In case  $W$  is a subspace of  $\mathbf{R}^m$  spanned by all column vectors  $\vec{u}_1, \dots, \vec{u}_n$  of an  $m \times n$  matrix  $A$ , then the orthogonal complement of  $W$  is the subspace

$$\begin{aligned} W^\perp &= \text{span}\{\vec{u}_1, \dots, \vec{u}_n\}^\perp \\ &= \{ \vec{y} : \vec{y} \cdot \vec{u}_i = 0 \text{ for all } i = 1, \dots, n \} \\ &= \text{nullspace}(A^T) \\ &= \text{span}\{\text{Strang's Special Solutions for } A^T \vec{u} = \vec{0}\} \end{aligned}$$

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**Method.** To compute a basis for  $W^\perp$ , find Strang's special Solutions for the homogeneous problem  $A^T \vec{u} = \vec{0}$ . The basis size is  $k = \text{number of free variables in } A^T \vec{u} = \vec{0}$ .

Applications may add an additional step to replace this basis by the Gram-Schmidt orthogonal basis  $\vec{y}_1, \dots, \vec{y}_k$ . Then  $W^\perp = \text{span}\{\vec{y}_1, \dots, \vec{y}_k\}$ .

## Fundamental Theorem of Linear Algebra

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**Definition.** The four fundamental subspaces are  $\text{rowspace}(\mathbf{A})$ ,  $\text{colspace}(\mathbf{A})$ ,  $\text{nullspace}(\mathbf{A})$  and  $\text{nullspace}(\mathbf{A}^T)$ .

The **Fundamental Theorem of Linear Algebra** has two parts:

(1) Dimension of the Four Fundamental Subspaces.

Assume matrix  $\mathbf{A}$  is  $m \times n$  with  $r$  pivots. Then

$$\begin{aligned}\dim(\text{rowspace}(\mathbf{A})) &= r, & \dim(\text{colspace}(\mathbf{A})) &= r, \\ \dim(\text{nullspace}(\mathbf{A})) &= n - r, & \dim(\text{nullspace}(\mathbf{A}^T)) &= m - r\end{aligned}$$

(2) Orthogonality of the Four Fundamental Subspaces.

$$\begin{aligned}\text{rowspace}(\mathbf{A}) &\perp \text{nullspace}(\mathbf{A}) \\ \text{colspace}(\mathbf{A}) &\perp \text{nullspace}(\mathbf{A}^T)\end{aligned}$$

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Gilbert Strang's textbook *Linear Algebra* has a cover illustration for the fundamental theorem of linear algebra. The original article is *The Fundamental Theorem of Linear Algebra*, <http://www.jstor.org/stable/2324660>. The free 1993 jstor PDF is available via the Marriott library. Requires UofU 2-factor login.