#### Orthogonal Complement

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Orthogonal Complement: The orthogonal complement of a subspace of V (or equivalently  $\mathbb{R}^n, \mathbb{C}^n$ ), W is given by

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Note: This is the set of all vectors orthogonal to all vectors in W.

#### Example

Using the standard inner product and  $V = \mathbb{R}^n$ . Let  $W = \operatorname{Span}\left(\begin{bmatrix}2\\4\\-6\end{bmatrix},\begin{bmatrix}0\\-1\\2\end{bmatrix}\right)$ , then  $W^{\perp} = \operatorname{Span}\left(\begin{bmatrix}-1\\2\\1\end{bmatrix}\right)$ 

## Computing Orthogonal Complements

#### **Theorem**

Let W be a subspace of V and A be a matrix such that  $W = \operatorname{Col}(A)$ . Then,

$$W^{\perp} = \operatorname{Nul}\left(A^{\top}\right)$$

Proof.

Let

$$A = \begin{bmatrix} \mathbf{v}_1 & \dots & \mathbf{v}_n \end{bmatrix}, W = \operatorname{\mathsf{Span}} (\mathbf{v}_1, \dots, \mathbf{v}_n)$$

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So,  $\mathbf{x} \in W^{\perp}$ . Thus, our two spaces are the same!

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- 4. Write out a basis of this nullspace

# Computing Orthogonal Set Example

Let's practice our algorithm!

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So, the null-space is given by the span of  $\begin{bmatrix} -1\\2\\1 \end{bmatrix}$ .

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A fundamental theorem behind much of linear algebra is how our "fundamental" subspaces  $(\operatorname{Col}(A), \operatorname{Row}(A), \operatorname{Nul}(A), \operatorname{Nul}(A^{\top}))$  relate to each other. It is summarized as

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## Orthogonal (Orthonormal) Sets

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#### Example

The set  $\left\{ \begin{bmatrix} 1\\1 \end{bmatrix}, \begin{bmatrix} -1\\1 \end{bmatrix} \right\}$  is orthogonal while  $\left\{ \frac{1}{\sqrt{2}} \begin{bmatrix} 1\\1 \end{bmatrix}, \frac{1}{\sqrt{2}} \begin{bmatrix} -1\\1 \end{bmatrix} \right\}$  is orthonormal

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We will show the equation

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has only the trivial solution  $c_1 = \cdots = c_m = 0$ .

We apply both sides of our equality to our inner product with  $\mathbf{u}_\ell$  for some  $1 \leq \ell \leq m$ , which gives us

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$$= c_{\ell} \langle \mathbf{u}_{\ell}, \mathbf{u}_{\ell} \rangle$$

Since we know that  $\mathbf{u}_{\ell}$  is non-zero,  $\langle \mathbf{u}_{\ell}, \mathbf{u}_{\ell} \rangle \neq 0$ , thus  $c_{\ell} = 0$ . Since  $\ell$  was some arbitrary index, all must be 0. Therefore, we have a list of linearly independent vectors.