

Lecture 13

Orthogonality

Inner product and orthogonality

Inner product and orthogonality

Recall we defined:

For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be .

Inner product and orthogonality

Recall we defined: For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be .

- $\langle \vec{x}, \vec{y} \rangle$ (or $\vec{x} \cdot \vec{y}$) = $\vec{x}^T \vec{y}$

Inner product and orthogonality

Recall we defined: For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be .

- $\boxed{\langle \vec{x}, \vec{y} \rangle \text{ (or } \vec{x} \cdot \vec{y}) = \vec{x}^T \vec{y}}$ inner product

Inner product and orthogonality

Recall we defined: For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be .

- $\boxed{\langle \vec{x}, \vec{y} \rangle \text{ (or } \vec{x} \cdot \vec{y}) = \vec{x}^T \vec{y}}$ inner product
- $\|\vec{x}\| = \langle \vec{x}, \vec{x} \rangle^{\frac{1}{2}}$

Inner product and orthogonality

Recall we defined: For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be .

- $\langle \vec{x}, \vec{y} \rangle$ (or $\vec{x} \cdot \vec{y}$) = $\vec{x}^T \vec{y}$ inner product
- $\|\vec{x}\| = \langle \vec{x}, \vec{x} \rangle^{\frac{1}{2}}$ length / norm.

Inner product and orthogonality

Recall we defined: For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be.

- $\langle \vec{x}, \vec{y} \rangle$ (or $\vec{x} \cdot \vec{y}$) = $\vec{x}^T \vec{y}$ inner product
- $\|\vec{x}\| = \sqrt{\langle \vec{x}, \vec{x} \rangle}$ length / norm.

e.g.: $\left\| \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\| = \sqrt{\langle \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \rangle} = \sqrt{\left([1 \ 2 \ 3] \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right)^T \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}} = \sqrt{1^2 + 2^2 + 3^2}$

Inner product and orthogonality

Recall we defined: For $\vec{x}, \vec{y} \in \mathbb{R}^n$, we defined their inner product to be.

- $\langle \vec{x}, \vec{y} \rangle$ (or $\vec{x} \cdot \vec{y}$) = $\vec{x}^T \vec{y}$ inner product
- $\|\vec{x}\| = \sqrt{\langle \vec{x}, \vec{x} \rangle}$ length / norm.

e.g.: $\left\| \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \right\| = \sqrt{\langle \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix} \rangle} = \sqrt{\begin{pmatrix} 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}^T} = \sqrt{1^2 + 2^2 + 3^2}$
is the distance from $\vec{0}$ to $\begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ in \mathbb{R}^3 .

Inner product and orthogonality

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

2. $\langle cx, y \rangle = (cx)^T y = c \cdot (x^T y) = c \langle x, y \rangle$.

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

3. $\langle x_1 + x_2, y \rangle = \langle x_1, y \rangle + \langle x_2, y \rangle.$

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle.$

2. $\langle cx, y \rangle = (cx)^T y = c \cdot (x^T y) = c \langle x, y \rangle.$

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

3. $\langle x_1 + x_2, y \rangle = \langle x_1, y \rangle + \langle x_2, y \rangle$.

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

2. $\langle cx, y \rangle = (cx)^T y = c \cdot (x^T y) = c \langle x, y \rangle$.

3. $\langle x_1 + x_2, y \rangle = (x_1 + x_2)^T y = x_1^T y + x_2^T y = \langle x_1, y \rangle + \langle x_2, y \rangle$

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

3. $\langle x_1 + x_2, y \rangle = \langle x_1, y \rangle + \langle x_2, y \rangle$.

4. $\langle x, x \rangle \geq 0$, and is zero iff $x = 0$.

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

2. $\langle cx, y \rangle = (cx)^T y = c \cdot (x^T y) = c \langle x, y \rangle$.

3. $\langle x_1 + x_2, y \rangle = (x_1 + x_2)^T y = x_1^T y + x_2^T y = \langle x_1, y \rangle + \langle x_2, y \rangle$

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

3. $\langle x_1 + x_2, y \rangle = \langle x_1, y \rangle + \langle x_2, y \rangle$.

4. $\langle x, x \rangle \geq 0$, and is zero iff $x = 0$.

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

2. $\langle cx, y \rangle = (cx)^T y = c \cdot (x^T y) = c \langle x, y \rangle$.

3. $\langle x_1 + x_2, y \rangle = (x_1 + x_2)^T y = x_1^T y + x_2^T y = \langle x_1, y \rangle + \langle x_2, y \rangle$

4. $\langle x, x \rangle = \sum_{i=1}^n (x_i)^2 \geq 0$

Inner product and orthogonality

Property of $\langle \cdot, \cdot \rangle$:

1. $\langle x, y \rangle = \langle y, x \rangle$

2. $\langle cx, y \rangle = c\langle x, y \rangle$

3. $\langle x_1 + x_2, y \rangle = \langle x_1, y \rangle + \langle x_2, y \rangle$.

4. $\langle x, x \rangle \geq 0$, and is zero iff $x = 0$.

Check: 1. $\langle x, y \rangle = x^T y = (x^T y)^T = y^T x = \langle y, x \rangle$.

2. $\langle cx, y \rangle = (cx)^T y = c \cdot (x^T y) = c \langle x, y \rangle$.

3. $\langle x_1 + x_2, y \rangle = (x_1 + x_2)^T y = x_1^T y + x_2^T y = \langle x_1, y \rangle + \langle x_2, y \rangle$

4. $\langle x, x \rangle = \sum_{i=1}^n (x_i)^2 \geq 0$, is zero iff $x_i = 0$ for all i
i.e. $x = 0$.

Inner product and orthogonality

Inner product and orthogonality

Geometric meaning:

For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

Inner product and orthogonality

Geometric meaning: For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

$$\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\| \|\vec{w}\|}$$

Inner product and orthogonality

Geometric meaning: For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

$$\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\| \|\vec{w}\|}$$

angle between two vectors

Inner product and orthogonality

Geometric meaning:

For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

$$\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\| \|\vec{w}\|}$$

angle between two vectors

Why?:

We see that angle is independent of the length of two vectors, assume $\|\vec{v}\| = 1 = \|\vec{w}\|$.

Inner product and orthogonality

Geometric meaning:

For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

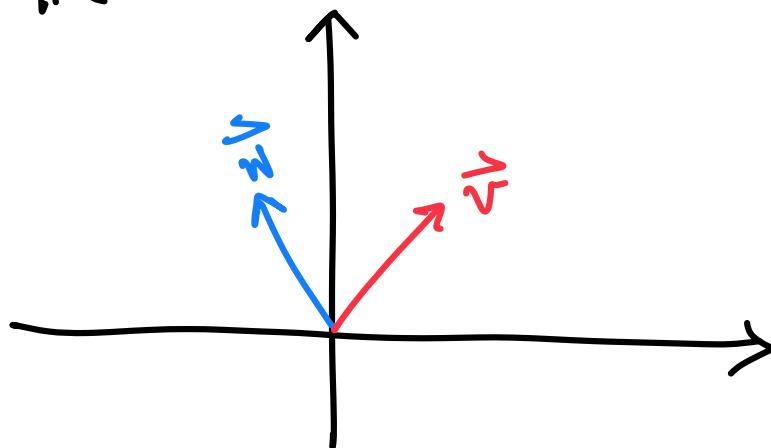
$$\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\| \|\vec{w}\|}$$

angle between two vectors

Why?:

We see that angle is independent of the length of two vectors, assume $\|\vec{v}\| = 1 = \|\vec{w}\|$.

\mathbb{R}^2 :



Inner product and orthogonality

Geometric meaning:

For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

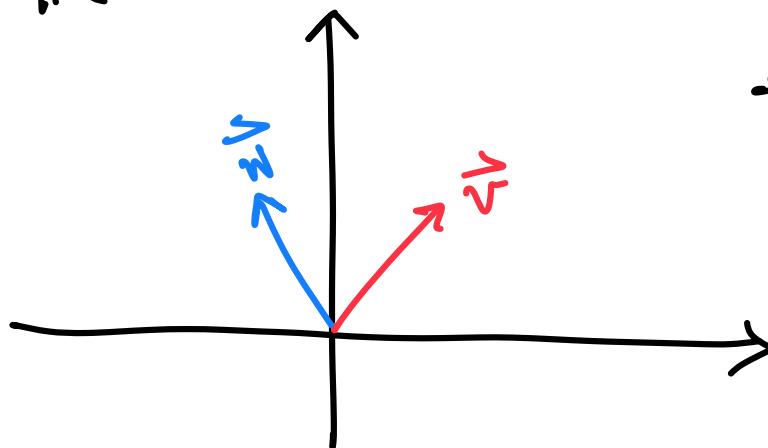
$$\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\| \|\vec{w}\|}$$

angle between two vectors

Why?:

We see that angle is independent of the length of two vectors, assume $\|\vec{v}\| = 1 = \|\vec{w}\|$.

\mathbb{R}^2 :



rotation
+ reflection : R.

Inner product and orthogonality

Geometric meaning:

For non-zero vectors $\vec{v}, \vec{w} \in \mathbb{R}^n$

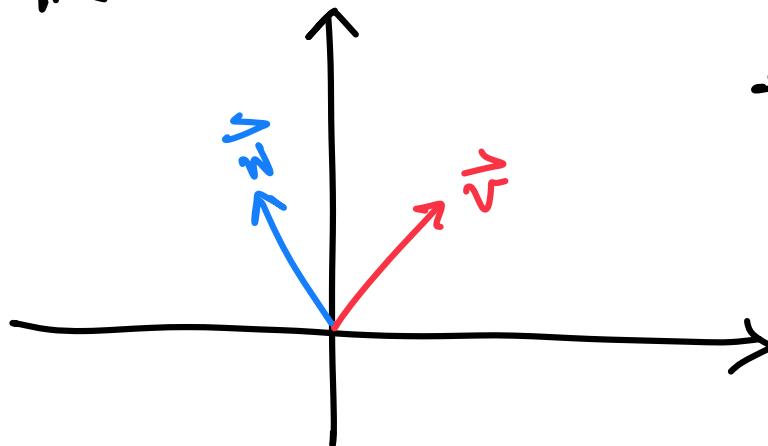
$$\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\| \|\vec{w}\|}$$

angle between two vectors

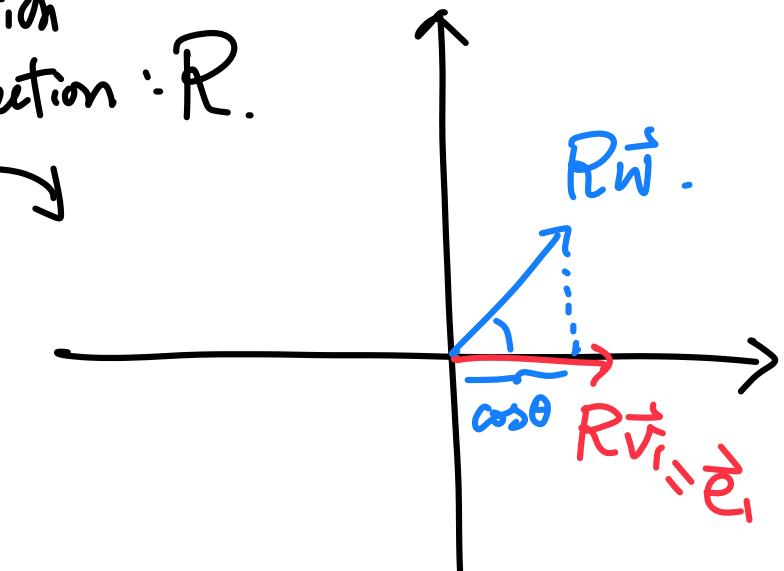
Why?:

We see that angle is independent of the length of two vectors, assume $\|\vec{v}\| = 1 = \|\vec{w}\|$.

\mathbb{R}^2 :



rotation
+ reflection : R .

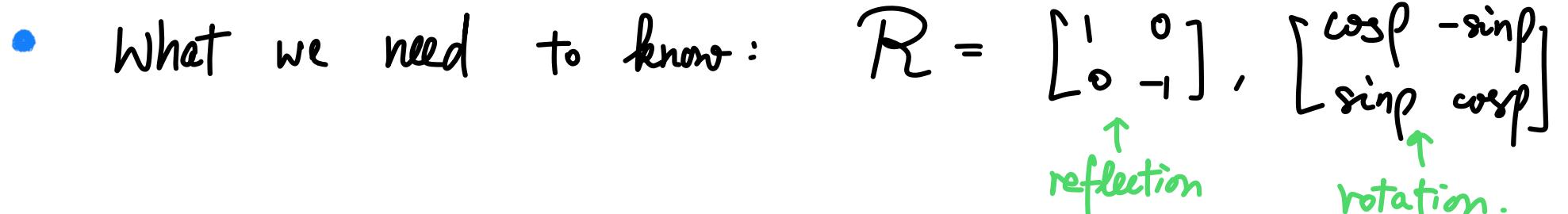


Inner product and orthogonality

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\phi & -\sin\phi \\ \sin\phi & \cos\phi \end{bmatrix}$

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$


The handwritten notes show the matrix $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}$ with a green arrow pointing to it labeled "reflection". Next to it is the matrix $\begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$ with a green arrow pointing to it labeled "rotation".

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\phi & -\sin\phi \\ \sin\phi & \cos\phi \end{bmatrix}$
 \uparrow
reflection rotation.

R preserve inner product: i.e. $\langle R\vec{x}, R\vec{y} \rangle = \langle \vec{x}, \vec{y} \rangle$

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$
 \uparrow
reflection rotation.

R preserve inner product: i.e. $\langle R\vec{x}, R\vec{y} \rangle_{\parallel} = \langle \vec{x}, \vec{y} \rangle$

$$\vec{x}^T R^T R \vec{y}$$

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\phi & -\sin\phi \\ \sin\phi & \cos\phi \end{bmatrix}$
 \uparrow
reflection rotation.

R preserve inner product: i.e. $\langle R\vec{x}, R\vec{y} \rangle_{\parallel} = \langle \vec{x}, \vec{y} \rangle$
 $\vec{x}^T R^T R \vec{y}$

- Notice $R^T R = I$ for both cases. $\Rightarrow R$ preserves inner product.

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$
 \uparrow
reflection rotation.

R preserve inner product: i.e. $\langle R\vec{x}, R\vec{y} \rangle_{\parallel} = \langle \vec{x}, \vec{y} \rangle$

$$\vec{x}^T R^T R \vec{y}$$

- Notice $R^T R = I$ for both cases. $\Rightarrow R$ preserves inner product.

- Any product of $\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$ still satisfies $R^T R = I$.

Inner product and orthogonality

- What we need to know: $R = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$
 \uparrow
reflection \uparrow
rotation.

R preserve inner product: i.e. $\langle R\vec{x}, R\vec{y} \rangle_{\parallel} = \langle \vec{x}, \vec{y} \rangle$
 $\vec{x}^T R^T R \vec{y}$

- Notice $R^T R = I$ for both cases. $\Rightarrow R$ preserves inner product.

- Any product of $\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix}, \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$ still satisfies $R^T R = I$.
- Any composition of rotations or reflection in \mathbb{R}^2 preserve $\langle \cdot, \cdot \rangle$.

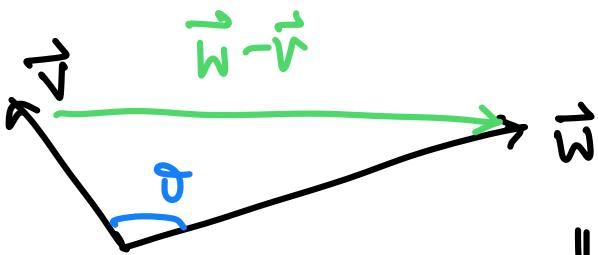
Inner product and orthogonality

Inner product and orthogonality

- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.

Inner product and orthogonality

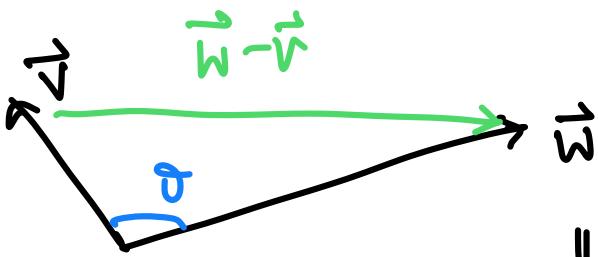
- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.



$$\|\vec{w} - \vec{v}\|^2 = \|\vec{v}\|^2 + \|\vec{w}\|^2 - 2\|\vec{v}\| \|\vec{w}\| \cos \theta.$$

Inner product and orthogonality

- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.

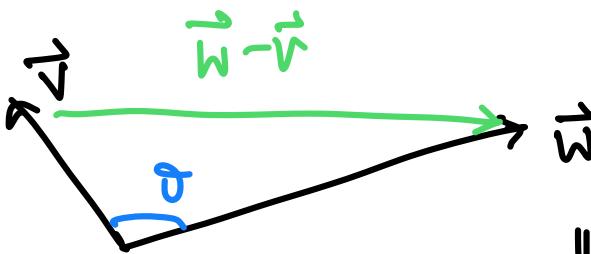


$$\|\vec{w} - \vec{v}\|^2 = \|\vec{v}\|^2 + \|\vec{w}\|^2 - 2\|\vec{v}\| \|\vec{w}\| \cos\theta.$$

- Expand the L.H.S.: $\|\vec{w} - \vec{v}\|^2 = (\vec{w} - \vec{v})^\top (\vec{w} - \vec{v})$

Inner product and orthogonality

- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.

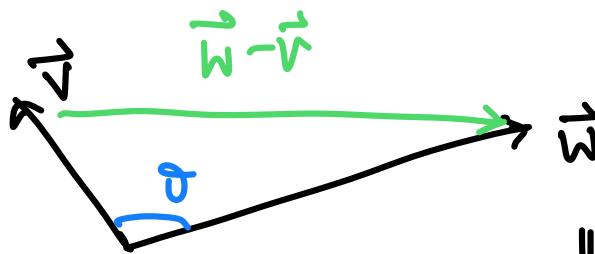


$$\|\vec{w}-\vec{v}\|^2 = \|\vec{v}\|^2 + \|\vec{w}\|^2 - 2\|\vec{v}\|\|\vec{w}\| \cos\theta.$$

- Expand the L.H.S.: $\|\vec{w}-\vec{v}\|^2 = (\vec{w}-\vec{v})^\top (\vec{w}-\vec{v})$
 $= \vec{w}^\top \vec{w} + \vec{v}^\top \vec{v} - \vec{v}^\top \vec{w} - \vec{w}^\top \vec{v}$

Inner product and orthogonality

- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.

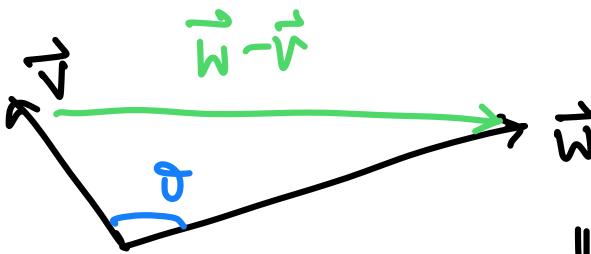


$$\|\vec{w}-\vec{v}\|^2 = \|\vec{v}\|^2 + \|\vec{w}\|^2 - 2\|\vec{v}\|\|\vec{w}\| \cos\theta.$$

- Expand the L.H.S.: $\|\vec{w}-\vec{v}\|^2 = (\vec{w}-\vec{v})^\top (\vec{w}-\vec{v})$
$$= \underbrace{\vec{w}^\top \vec{w}}_{\|\vec{w}\|^2} + \underbrace{\vec{v}^\top \vec{v}}_{\|\vec{v}\|^2} - \underbrace{\vec{v}^\top \vec{w}}_{\langle \vec{v}, \vec{w} \rangle} - \underbrace{\vec{w}^\top \vec{v}}_{\langle \vec{v}, \vec{w} \rangle}$$

Inner product and orthogonality

- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.

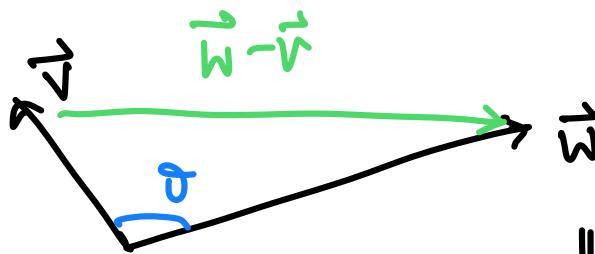


$$\|\vec{w} - \vec{v}\|^2 = \|\vec{v}\|^2 + \|\vec{w}\|^2 - 2\|\vec{v}\| \|\vec{w}\| \cos\theta.$$

- Expand the L.H.S.: $\|\vec{w} - \vec{v}\|^2 = (\vec{w} - \vec{v})^\top (\vec{w} - \vec{v})$
$$= \underbrace{\vec{w}^\top \vec{w}}_{\|\vec{w}\|^2} + \underbrace{\vec{v}^\top \vec{v}}_{\|\vec{v}\|^2} - \underbrace{\vec{v}^\top \vec{w}}_{\langle \vec{v}, \vec{w} \rangle} - \underbrace{\vec{w}^\top \vec{v}}_{\langle \vec{v}, \vec{w} \rangle}$$
$$= \|\vec{w}\|^2 + \|\vec{v}\|^2 - 2\langle \vec{v}, \vec{w} \rangle.$$

Inner product and orthogonality

- For general vectors \vec{v}, \vec{w} in \mathbb{R}^n : we have.
the cosine formula.



$$\|\vec{w} - \vec{v}\|^2 = \|\vec{v}\|^2 + \|\vec{w}\|^2 - 2\|\vec{v}\|\|\vec{w}\| \cos\theta.$$

- Expand the L.H.S.: $\|\vec{w} - \vec{v}\|^2 = (\vec{w} - \vec{v})^\top (\vec{w} - \vec{v})$
$$= \underbrace{\vec{w}^\top \vec{w}}_{\|\vec{w}\|^2} + \underbrace{\vec{v}^\top \vec{v}}_{\|\vec{v}\|^2} - \underbrace{\vec{v}^\top \vec{w}}_{\langle \vec{v}, \vec{w} \rangle} - \underbrace{\vec{w}^\top \vec{v}}_{\langle \vec{v}, \vec{w} \rangle}$$
$$= \|\vec{w}\|^2 + \|\vec{v}\|^2 - 2\langle \vec{v}, \vec{w} \rangle.$$
- Comparing we have $\cos\theta = \frac{\langle \vec{v}, \vec{w} \rangle}{\|\vec{v}\|\|\vec{w}\|}.$

orthogonality

orthogonality

Def:

For two vectors \vec{x}, \vec{y} in \mathbb{R}^n , we say

orthogonality

Def:

- For two vectors \vec{x}, \vec{y} in \mathbb{R}^n , we say
- They are orthogonal (正交) if $\langle \vec{x}, \vec{y} \rangle = 0$

orthogonality

Def:

For two vectors \vec{x}, \vec{y} in \mathbb{R}^n , we say

- They are orthogonal (正交) if $\langle \vec{x}, \vec{y} \rangle = 0$
- Angle between them is less than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle > 0$

orthogonality

Def:

For two vectors \vec{x}, \vec{y} in \mathbb{R}^n , we say

- They are orthogonal (正交) if $\langle \vec{x}, \vec{y} \rangle = 0$
- Angle between them is less than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle > 0$
- " Greater than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle < 0$.

orthogonality

Def.

For two vectors \vec{x}, \vec{y} in \mathbb{R}^n , we say

- They are orthogonal (正交) if $\langle \vec{x}, \vec{y} \rangle = 0$
- Angle between them is less than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle > 0$
- " greater than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle < 0$.

Def.

Two subspace W_1, W_2 of \mathbb{R}^n are said to be orthogonal if $\langle \vec{w}_1, \vec{w}_2 \rangle = 0$ for every $\vec{w}_1 \in W_1, \vec{w}_2 \in W_2$.
We write $W_1 \perp W_2$.

orthogonality

Def.

For two vectors \vec{x}, \vec{y} in \mathbb{R}^n , we say

- They are orthogonal (正交) if $\langle \vec{x}, \vec{y} \rangle = 0$
- Angle between them is less than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle > 0$
- " greater than $\frac{\pi}{2}$ if $\langle \vec{x}, \vec{y} \rangle < 0$.

Def.

Two subspace W_1, W_2 of \mathbb{R}^n are said to be orthogonal if $\langle \vec{w}_1, \vec{w}_2 \rangle = 0$ for every $\vec{w}_1 \in W_1, \vec{w}_2 \in W_2$.
We write $W_1 \perp W_2$.

e.g.

$\vec{x} = \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}, \vec{y} = \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix}$ are orthogonal.

orthogonality

orthogonality

Prop:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n
s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

orthogonality

Prop:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n
s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

then $\vec{v}_1, \dots, \vec{v}_k$ are l.i.

orthogonality

Pmp:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n

s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

then $\vec{v}_1, \dots, \vec{v}_k$ are l.i.

Pf:

Assuming $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = 0$.

orthogonality

Pmp:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n
s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

then $\vec{v}_1, \dots, \vec{v}_k$ are l.i.

Pf:

Assuming $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = 0$.

then $\langle \vec{v}_i, c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k \rangle = 0$

orthogonality

Pmp:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n
s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

then $\vec{v}_1, \dots, \vec{v}_k$ are l.i.

Pf:

Assuming $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = 0$.

then $\langle \vec{v}_i, c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k \rangle = 0$

$\Rightarrow c_i \langle \vec{v}_i, \vec{v}_i \rangle = 0$.

orthogonality

Pmp:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n
s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

then $\vec{v}_1, \dots, \vec{v}_k$ are l.i.

Pf:

Assuming $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = 0$.

then $\langle \vec{v}_i, c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k \rangle = 0$

\Rightarrow

$$c_i \langle \vec{v}_i, \vec{v}_i \rangle = 0 .$$

$\neq 0$.

orthogonality

Pmp:

If $\vec{v}_1, \dots, \vec{v}_k$ are non-zero vectors in \mathbb{R}^n
s.t. $\langle \vec{v}_i, \vec{v}_j \rangle = 0$ for $i \neq j$.

then $\vec{v}_1, \dots, \vec{v}_k$ are l.i.

Pf:

Assuming $c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k = 0$.

then $\langle \vec{v}_i, c_1\vec{v}_1 + c_2\vec{v}_2 + \dots + c_k\vec{v}_k \rangle = 0$

\Rightarrow

$$c_i \langle \vec{v}_i, \vec{v}_i \rangle = 0.$$

$$\Rightarrow c_i = 0$$

This is for any
 $i = 1, \dots, k$.

orthogonality

orthogonality

Recall that :

We look at $m \times n$ matrix A with
associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$

$$C(A^T) \perp N(A), C(A^T) + N(A) = \mathbb{R}^n.$$

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$

$$C(A^T) \perp N(A), C(A^T) + N(A) = \mathbb{R}^n.$$

Recall the proof:

Say $C(A^T) \perp N(A)$.

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$

$$C(A^T) \perp N(A), C(A^T) + N(A) = \mathbb{R}^n.$$

Recall the proof:

Say $C(A^T) \perp N(A)$.

- take $\vec{x} \in C(A^T)$, $\vec{y} \in N(A)$, want $\langle \vec{x}, \vec{y} \rangle = 0$

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$
 $C(A^T) \perp N(A)$, $C(A^T) + N(A) = \mathbb{R}^n$.

Recall the proof:

Say $C(A^T) \perp N(A)$.

- take $\underbrace{\vec{x}}_{\in C(A^T)} \in C(A^T)$, $\underbrace{\vec{y}}_{\in N(A)} \in N(A)$, want $\langle \vec{x}, \vec{y} \rangle = 0$
- $$\vec{x} = A^T \vec{z}$$
- $$A \vec{y} = 0.$$

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$
 $C(A^T) \perp N(A)$, $C(A^T) + N(A) = \mathbb{R}^n$.

Recall the proof:

Say $C(A^T) \perp N(A)$.

- take $\underbrace{\vec{x}}_{\in C(A^T)} \in C(A^T)$, $\underbrace{\vec{y}}_{\in N(A)} \in N(A)$, want $\langle \vec{x}, \vec{y} \rangle = 0$
- $$\vec{x} = A^T \vec{z} \in \mathbb{R}^m. \quad A\vec{y} = 0.$$

orthogonality

Recall that :

We look at $m \times n$ matrix A with associated $L_A : \mathbb{R}^n \longrightarrow \mathbb{R}^m$

- We learn that $C(A) \perp N(A^T)$, $C(A) + N(A^T) = \mathbb{R}^m$
 $C(A^T) \perp N(A)$, $C(A^T) + N(A) = \mathbb{R}^n$.

Recall the proof:

Say $C(A^T) \perp N(A)$.

- take $\underbrace{\vec{x}}_{\in C(A^T)} \in C(A^T)$, $\underbrace{\vec{y}}_{\in N(A)} \in N(A)$, want $\langle \vec{x}, \vec{y} \rangle = 0$
- $\vec{x} = A^T \vec{z} \in \mathbb{R}^m$. $A\vec{y} = 0$.
- $\langle \vec{x}, \vec{y} \rangle = \langle A^T \vec{z}, \vec{y} \rangle = (A^T \vec{z})^T \cdot \vec{y} = \vec{z}^T A \vec{y} = 0$

orthogonality

orthogonality

Recall:

How we get $C(A^T) + N(A) = \mathbb{R}^n$?

orthogonality

Recall: How we get $C(A^T) + N(A) = \mathbb{R}^n$?

- $\dim(C(A^T)) = \text{rk}(A) = r, \quad \dim(N(A)) = n - r.$

orthogonality

Recall: How we get $C(A^T) + N(A) = \mathbb{R}^n$?

- $\dim(C(A^T)) = \text{rk}(A) = r, \quad \dim(N(A)) = n - r.$
- $C(A^T) \cap N(A) = \{0\}.$

orthogonality

Recall: How we get $C(A^T) + N(A) = \mathbb{R}^n$?

- $\dim(C(A^T)) = \text{rk}(A) = r, \quad \dim(N(A)) = n - r.$
- $C(A^T) \cap N(A) = \{0\}.$ fact: $W_1 \perp W_2$ then
 $W_1 \cap W_2 = \{0\}.$

orthogonality

Recall: How we get $C(A^T) + N(A) = \mathbb{R}^n$?

- $\dim(C(A^T)) = \text{rk}(A) = r, \quad \dim(N(A)) = n - r.$
- $C(A^T) \cap N(A) = \{0\}.$ fact: $W_1 \perp W_2$ then
 $W_1 \cap W_2 = \{0\}.$
- Fix a basis $\vec{v}_1, \dots, \vec{v}_r$ for $C(A^T)$
 $\vec{w}_1, \dots, \vec{w}_{n-r}$ for $N(A).$

orthogonality

Recall: How we get $C(A^T) + N(A) = \mathbb{R}^n$?

- $\dim(C(A^T)) = \text{rk}(A) = r, \quad \dim(N(A)) = n - r.$
- $C(A^T) \cap N(A) = \{0\}.$ fact: $W_1 \perp W_2$ then
 $W_1 \cap W_2 = \{0\}.$
- Fix a basis $\vec{v}_1, \dots, \vec{v}_r$ for $C(A^T)$
 $\vec{w}_1, \dots, \vec{w}_{n-r}$ for $N(A).$
 $\Rightarrow \vec{v}_1, \vec{v}_2, \dots, \vec{v}_r, \vec{w}_1, \vec{w}_2, \dots, \vec{w}_{n-r}$ a basis for
 $C(A^T) + N(A)$

orthogonality

Recall: How we get $C(A^T) + N(A) = \mathbb{R}^n$?

- $\dim(C(A^T)) = \text{rk}(A) = r, \dim(N(A)) = n - r.$
- $C(A^T) \cap N(A) = \{0\}$. fact: $W_1 \perp W_2$ then
 $W_1 \cap W_2 = \{0\}$.
- Fix a basis $\vec{v}_1, \dots, \vec{v}_r$ for $C(A^T)$
 $\vec{w}_1, \dots, \vec{w}_{n-r}$ for $N(A)$.
 $\Rightarrow \vec{v}_1, \vec{v}_2, \dots, \vec{v}_r, \vec{w}_1, \vec{w}_2, \dots, \vec{w}_{n-r}$ a basis for
 $C(A^T) + N(A)$

fact: If $W_1 \cap W_2 = \{0\}$, then a basis $\vec{v}_1, \dots, \vec{v}_k$ for W_1
 $\vec{w}_1, \dots, \vec{w}_\ell$ for W_2 .
 $\vec{v}_1, \dots, \vec{v}_k, \vec{w}_1, \dots, \vec{w}_\ell$ is a basis for $W_1 + W_2$.

orthogonality

orthogonality

- $\vec{v}_1, \dots, \vec{v}_r, \vec{w}_1, \dots, \vec{w}_{n-r}$ is a basis for \mathbb{R}^n .

orthogonality

- $\vec{v}_1, \dots, \vec{v}_r, \vec{w}_1, \dots, \vec{w}_{n-r}$ is a basis for \mathbb{R}^n .

fact: 1. If $\dim(V) = n$, $\vec{v}_1, \dots, \vec{v}_n$ l.i \Rightarrow basis

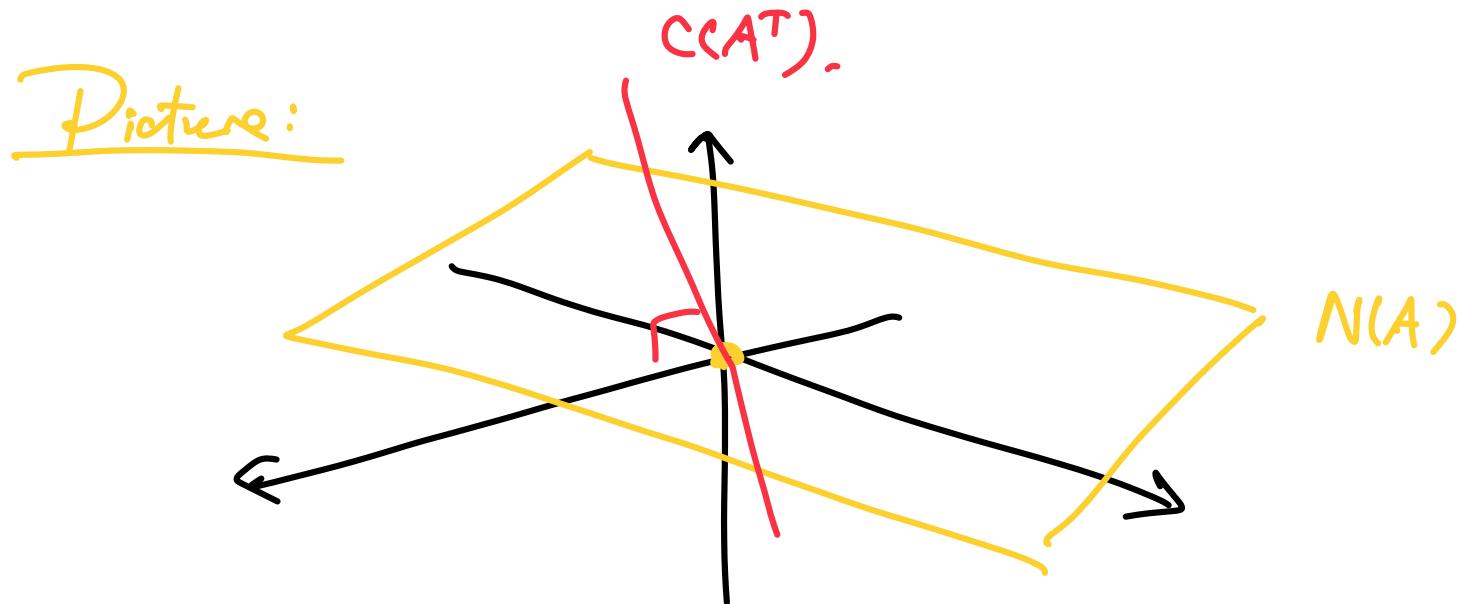
2. If $\dim(V) = n$, $\text{Span}\{\vec{v}_1, \dots, \vec{v}_n\} = V \Rightarrow$ basis.

orthogonality

- $\vec{v}_1, \dots, \vec{v}_r, \vec{w}_1, \dots, \vec{w}_{n-r}$ is a basis for \mathbb{R}^n .
fact: 1. If $\dim(V) = n$, $\vec{v}_1, \dots, \vec{v}_n$ l.i \Rightarrow basis
2. If $\dim(V) = n$, $\text{Span}\{\vec{v}_1, \dots, \vec{v}_n\} = V \Rightarrow$ basis.
- $C(A^\top) + N(A) = \mathbb{R}^n$.

orthogonality

- $\vec{v}_1, \dots, \vec{v}_r, \vec{w}_1, \dots, \vec{w}_{n-r}$ is a basis for \mathbb{R}^n .
- fact: 1. If $\dim(V) = n$, $\vec{v}_1, \dots, \vec{v}_n$ l.i \Rightarrow basis
- 2. If $\dim(V) = n$, $\text{Span}\{\vec{v}_1, \dots, \vec{v}_n\} = V \Rightarrow$ basis.
- $C(A^T) + N(A) = \mathbb{R}^n$.



orthogonality complement

orthogonality complement

Def:

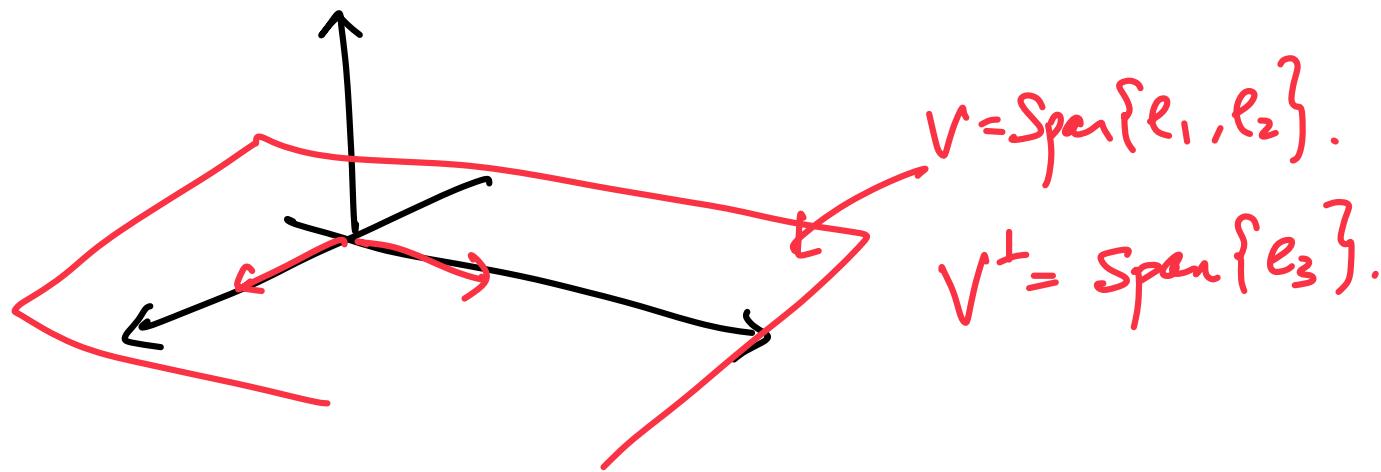
$V \subseteq \mathbb{R}^n$ a subspace, then we define the orthogonal complement of V , denoted by $V^\perp = \left\{ \vec{w} \in \mathbb{R}^n \mid \langle \vec{v}, \vec{w} \rangle = 0 \text{ for all } \vec{v} \in V \right\}$

orthogonality complement

Def:

$V \subseteq \mathbb{R}^n$ a subspace, then we define the orthogonal complement of V , denoted by $V^\perp = \left\{ \vec{w} \in \mathbb{R}^n \mid \langle \vec{v}, \vec{w} \rangle = 0 \text{ for all } \vec{v} \in V \right\}$

e.g.

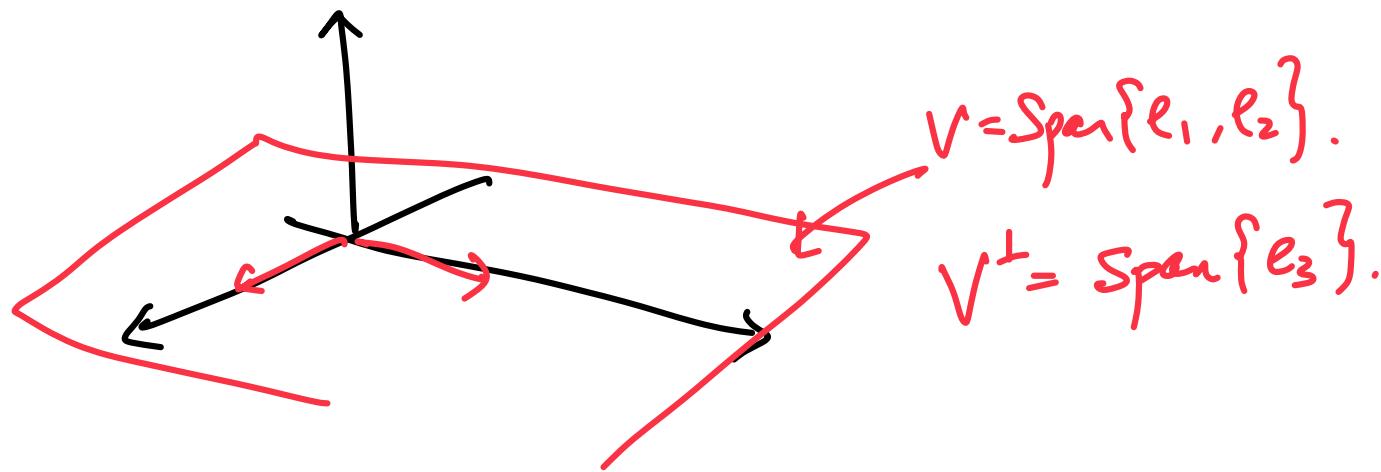


orthogonality complement

Def:

$V \subseteq \mathbb{R}^n$ a subspace, then we define the orthogonal complement of V , denoted by $V^\perp = \left\{ \vec{w} \in \mathbb{R}^n \mid \langle \vec{v}, \vec{w} \rangle = 0 \text{ for all } \vec{v} \in V \right\}$

e.g.



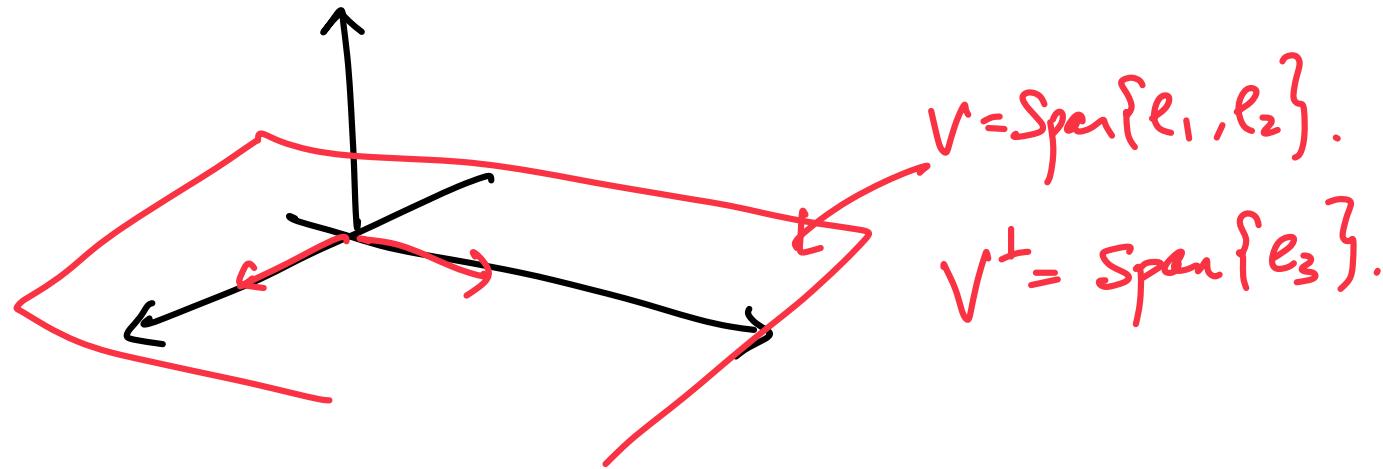
Prop: V^\perp is a subspace, $V^\perp \cap V = \{0\}$, $V^\perp + V = \mathbb{R}^n$

orthogonality complement

Def:

$V \subseteq \mathbb{R}^n$ a subspace, then we define the orthogonal complement of V , denoted by $V^\perp = \left\{ \vec{w} \in \mathbb{R}^n \mid \langle \vec{v}, \vec{w} \rangle = 0 \text{ for all } \vec{v} \in V \right\}$

e.g.



Prop: V^\perp is a subspace, $V^\perp \cap V = \{0\}$, $V^\perp + V = \mathbb{R}^n$

Pf: • if $w_1, w_2 \in V^\perp$

$$\langle c_1 \vec{w}_1 + c_2 \vec{w}_2, \vec{v} \rangle = c_1 \langle \vec{w}_1, \vec{v} \rangle + c_2 \langle \vec{w}_2, \vec{v} \rangle = 0 \quad \text{for all } v \in V$$

orthogonality complement

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have
$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$
then $\vec{v} = 0$.
- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional
take $\vec{v}_1, \dots, \vec{v}_r$ in V be a basis.

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional

take $\vec{v}_1, \dots, \vec{v}_r$ in V be a basis.

$$\rightsquigarrow A := \begin{bmatrix} -\vec{v}_1^T \\ -\vec{v}_2^T \\ \vdots \\ -\vec{v}_r^T \end{bmatrix},$$

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional

take $\vec{v}_1, \dots, \vec{v}_r$ in V be a basis.

$$\rightsquigarrow A := \begin{bmatrix} -\vec{v}_1^T \\ -\vec{v}_2^T \\ \vdots \\ -\vec{v}_r^T \end{bmatrix}, \quad V^\perp = N(A).$$

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional

take $\vec{v}_1, \dots, \vec{v}_r$ in V be a basis.

$$\rightsquigarrow A := \begin{bmatrix} -\vec{v}_1^T \\ -\vec{v}_2^T \\ \vdots \\ -\vec{v}_r^T \end{bmatrix}, \quad V^\perp = N(A).$$

fact: if $\langle \vec{w}, \vec{v}_i \rangle = 0$ for a basis $\vec{v}_1, \dots, \vec{v}_k$ of V
then $\langle \vec{w}, \vec{v} \rangle = 0$ for all $\vec{v} \in V$.

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional

take $\vec{v}_1, \dots, \vec{v}_r$ in V be a basis.

$$\rightsquigarrow A := \begin{bmatrix} -\vec{v}_1^T \\ -\vec{v}_2^T \\ \vdots \\ -\vec{v}_r^T \end{bmatrix}, \quad V^\perp = N(A).$$

$$\begin{cases} \text{rk}(A) = r \\ \dim N(A) = n - r \end{cases}$$

fact: if $\langle \vec{w}, \vec{v}_i \rangle = 0$ for a basis $\vec{v}_1, \dots, \vec{v}_k$ of V
then $\langle \vec{w}, \vec{v} \rangle = 0$ for all $\vec{v} \in V$.

orthogonality complement

- $V \cap V^\perp = \{0\}$: take $\vec{v} \in V \cap V^\perp$ then we have

$$\langle \vec{v}, \vec{v} \rangle = \|\vec{v}\|^2 = 0$$

then $\vec{v} = 0$.

- $V + V^\perp = \mathbb{R}^n$: Say V is r -dimensional

take $\vec{v}_1, \dots, \vec{v}_r$ in V be a basis.

$$\rightsquigarrow A := \begin{bmatrix} -\vec{v}_1^T \\ -\vec{v}_2^T \\ \vdots \\ -\vec{v}_r^T \end{bmatrix}, \quad V^\perp = N(A).$$

$\dim N(A) = n-r$
 $\text{rk}(A) = r$

↑ $n-r$ dimensional.

fact: if $\langle \vec{w}, \vec{v}_i \rangle = 0$ for a basis $\vec{v}_1, \dots, \vec{v}_k$ of V
then $\langle \vec{w}, \vec{v} \rangle = 0$ for all $\vec{v} \in V$.

orthogonality complement

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write
$$\vec{v} = \vec{v}_1 + \vec{v}_2.$$

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \vec{v}_1 + \vec{v}_2$$

\vec{v} \vec{v}_1 \vec{v}_2

\vec{v} \vec{v}^\perp

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \underbrace{\vec{v}_1}_{v} + \underbrace{\vec{v}_2}_{v^\perp}$$

this decomposition is unique.

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \vec{v}_1 + \vec{v}_2$$

\vec{v} \vec{v}_1 \vec{v}_2

\vec{v} \vec{v}^\perp

this decomposition is unique.

- $(\vec{v}^\perp)^\perp = \vec{v}$

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \vec{v}_1 + \vec{v}_2$$

\vec{v} \vec{v}^\perp

this decomposition is unique.

- $(V^\perp)^\perp = V$

Reason: $V \subseteq (V^\perp)^\perp$ by definition. and

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \underbrace{\vec{v}_1}_{V} + \underbrace{\vec{v}_2}_{V^\perp}$$

this decomposition is unique.

- $(V^\perp)^\perp = V$

Reason: $V \subseteq (V^\perp)^\perp$ by definition. and

$$V + V^\perp = \mathbb{R}^n = V^\perp + (V^\perp)^\perp$$

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \vec{v}_1 + \vec{v}_2$$

\vec{v} \vec{v}^\perp

this decomposition is unique.

- $(V^\perp)^\perp = V$

Reason: $V \subseteq (V^\perp)^\perp$ by definition. and

$$V + V^\perp = \mathbb{R}^n = V^\perp + (V^\perp)^\perp$$
$$\Rightarrow \dim V = \dim (V^\perp)^\perp$$

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \underbrace{\vec{v}_1}_{V} + \underbrace{\vec{v}_2}_{V^\perp}$$

this decomposition is unique.

- $(V^\perp)^\perp = V$

Reason: $V \subseteq (V^\perp)^\perp$ by definition. and

$$V + V^\perp = \mathbb{R}^n = V^\perp + (V^\perp)^\perp$$
$$\Rightarrow \dim V = \dim (V^\perp)^\perp \Rightarrow V = (V^\perp)^\perp$$

orthogonality complement

- The theorem say given $\vec{v} \in \mathbb{R}^n$, we can write

$$\vec{v} = \underbrace{\vec{v}_1}_{V} + \underbrace{\vec{v}_2}_{V^\perp}$$

this decomposition is unique.

- $(V^\perp)^\perp = V$

Reason: $V \subseteq (V^\perp)^\perp$ by definition. and

$$V + V^\perp = \mathbb{R}^n = V^\perp + (V^\perp)^\perp$$

$$\Rightarrow \dim V = \dim (V^\perp)^\perp \Rightarrow V = (V^\perp)^\perp$$

fact: if $W \subseteq V$ be a subspace, $\dim W = \dim V$
 $\Rightarrow W = V$.

orthogonality complement

orthogonality complement

- Therefore we may write $C(A^T)^\perp = N(A)$, $C(A)^\perp = N(A^T)$.
i.e. they are orthogonal complement.

orthogonality complement

- Therefore we may write $C(A^T)^\perp = N(A)$, $C(A)^\perp = N(A^T)$.
i.e. they are orthogonal complement.

Another way to read this:

orthogonality complement

- Therefore we may write $C(A^T)^\perp = N(A)$, $C(A)^\perp = N(A^T)$.
i.e. they are orthogonal complements.

Another way to read this:

Thm: $Ax = b$ is solvable iff $b^T y = 0$ for all $y \in N(A^T)$

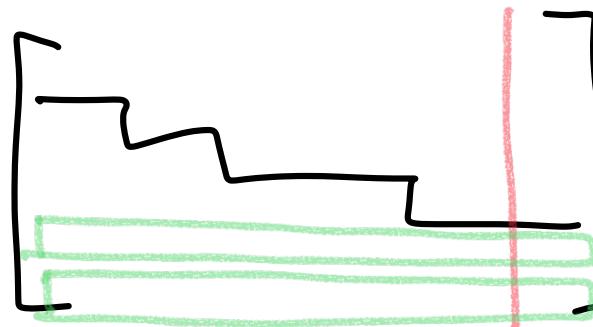
orthogonality complement

- Therefore we may write $C(A^T)^\perp = N(A)$, $C(A)^\perp = N(A^T)$.
i.e. they are orthogonal complements.

Another way to read this:

Thm: $Ax = b$ is solvable iff $b^T y = 0$ for all $y \in N(A^T)$

Recall $[A|b] \xrightarrow{\text{Gauss E.}}$



orthogonality complement

- Therefore we may write $C(A^T)^\perp = N(A)$, $C(A)^\perp = N(A^T)$.
i.e. they are orthogonal complements.

Another way to read this:

Thm: $Ax = b$ is solvable iff $b^T y = 0$ for all $y \in N(A^T)$

Recall $[A|b] \xrightarrow{\text{Gauss E.}}$

This gives
equation
on b in $C(A)$.

orthogonality complement

- Therefore we may write $C(A^T)^\perp = N(A)$, $C(A)^\perp = N(A^T)$.
i.e. they are orthogonal complements.

Another way to read this:

Thm: $Ax = b$ is solvable iff $b^T y = 0$ for all $y \in N(A^T)$

Recall $[A|b] \xrightarrow{\text{Gauss E.}}$

The diagram shows a square matrix in row echelon form. The first three columns are highlighted with a green horizontal bar. A vertical red line is drawn through the fourth column. Red arrows point from the text "This gives equation on b in C(A)" to the green bar and the vertical red line.

- Say $b_1 + 2b_2 + 5b_3 = 0$. is one of the equations.
 $[1, 2, 5]^T$ is a vector in $N(A^T)$.

orthogonality complement

orthogonality complement

Picture:

$\mathcal{R}^m:$

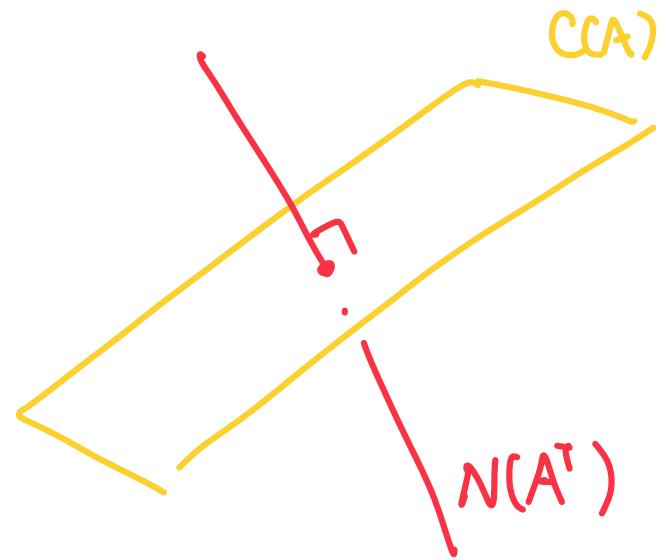
$\mathcal{D}^n:$

orthogonality complement

Picture:

$\mathcal{R}^m:$

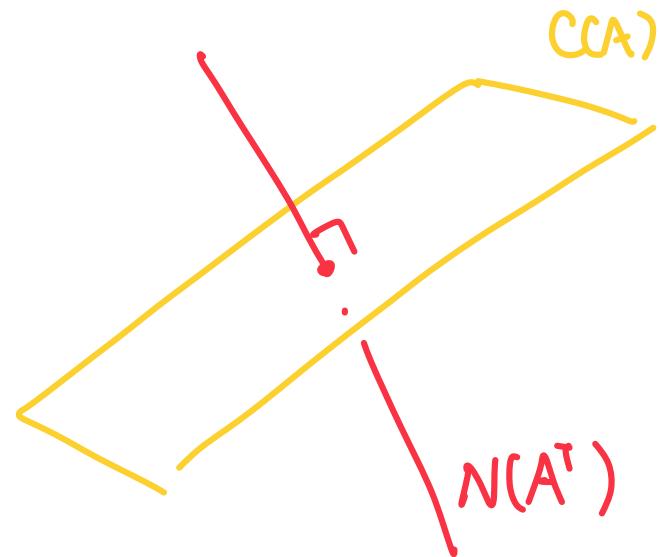
$\mathcal{D}^n:$



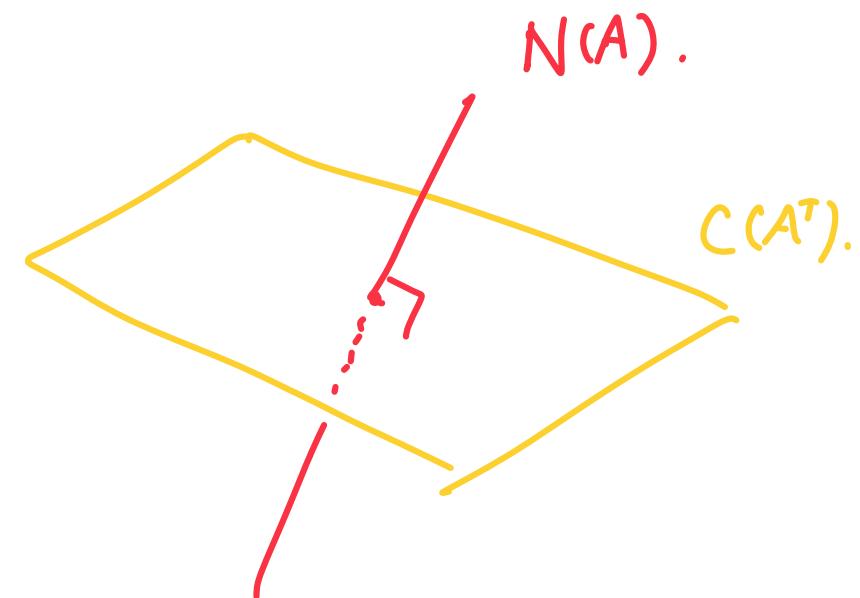
orthogonality complement

Picture:

\mathbb{R}^m :



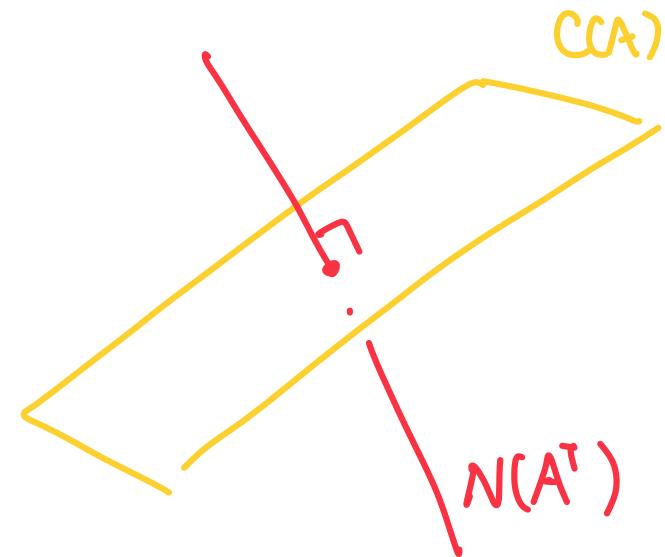
\mathbb{R}^n :



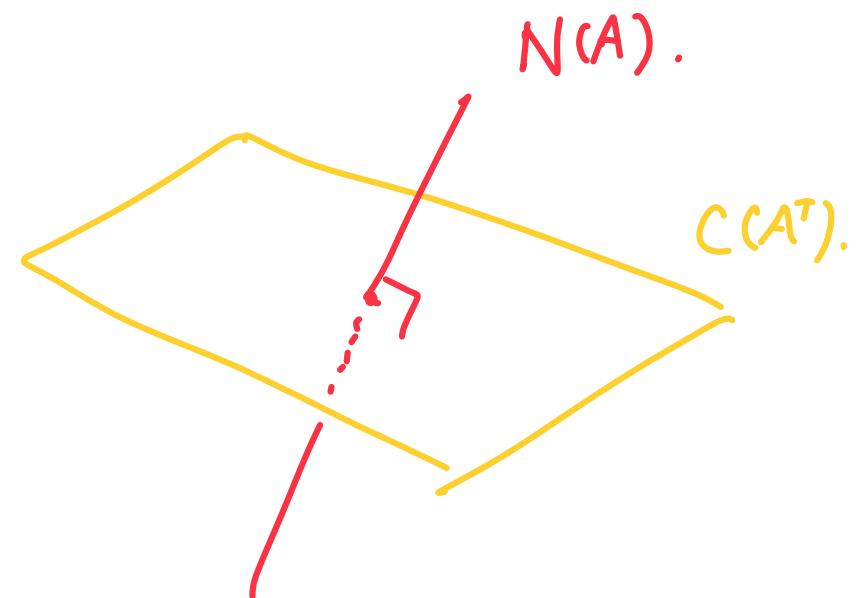
orthogonality complement

Picture:

\mathbb{R}^m :



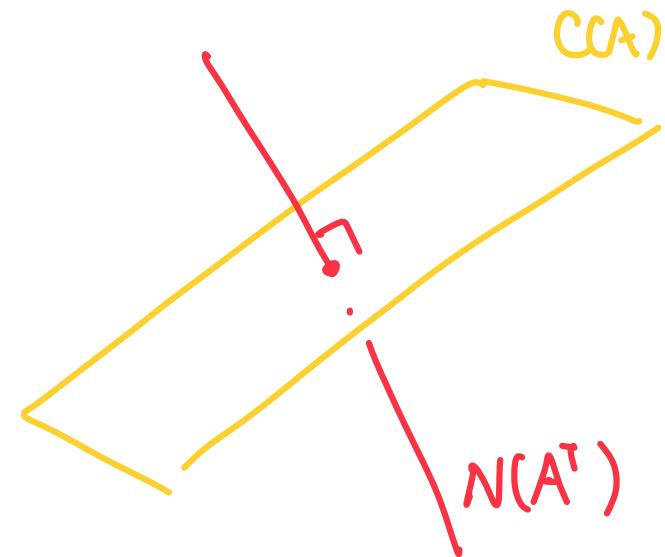
\mathbb{R}^n :



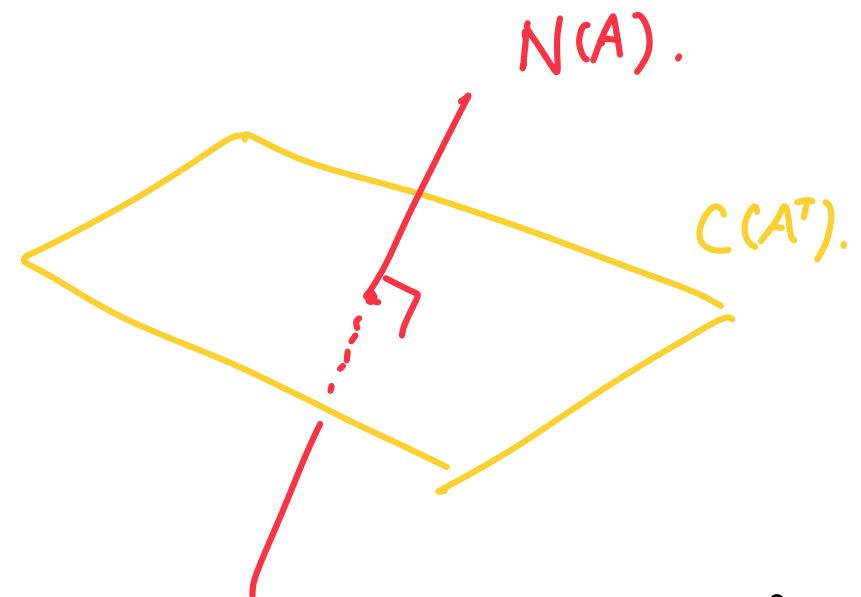
orthogonality complement

Picture:

\mathbb{R}^m :



\mathbb{R}^n :

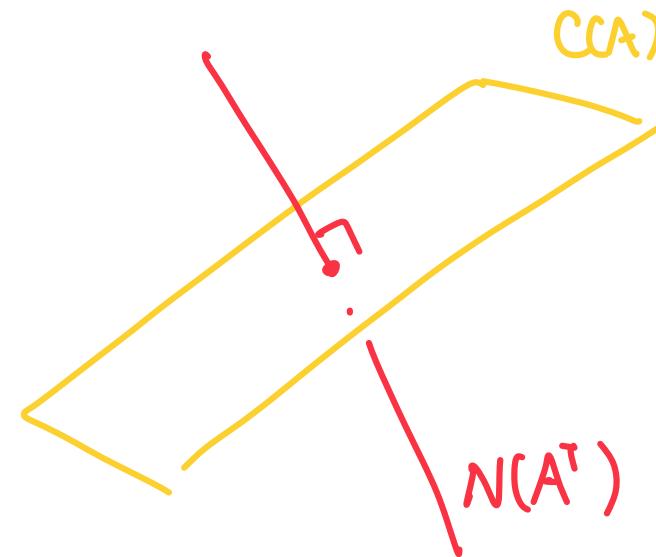


- L_A send $N(A)$ to $\{\vec{0}\}$, L_{A^\top} send $N(A^\top)$ to $\{\vec{0}\}$.

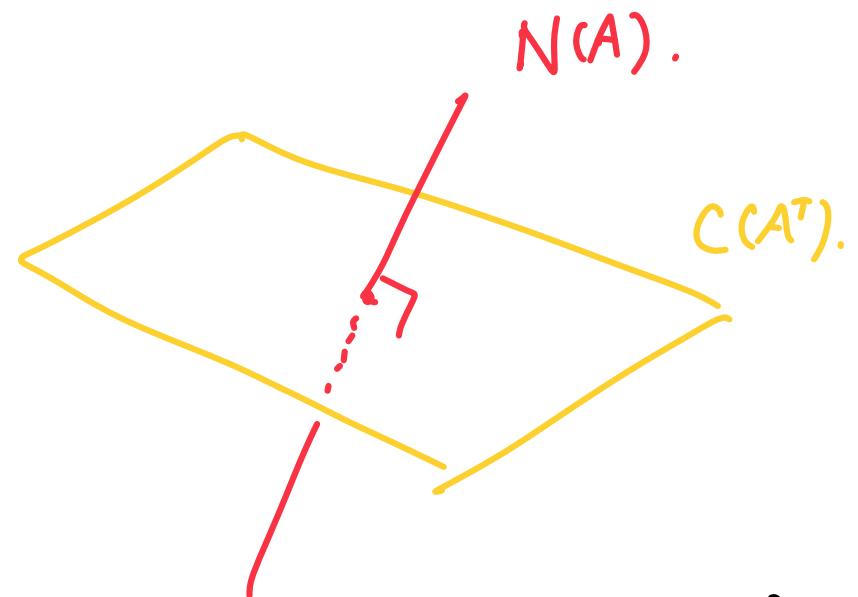
orthogonality complement

Picture:

\mathbb{R}^m :



\mathbb{R}^n :

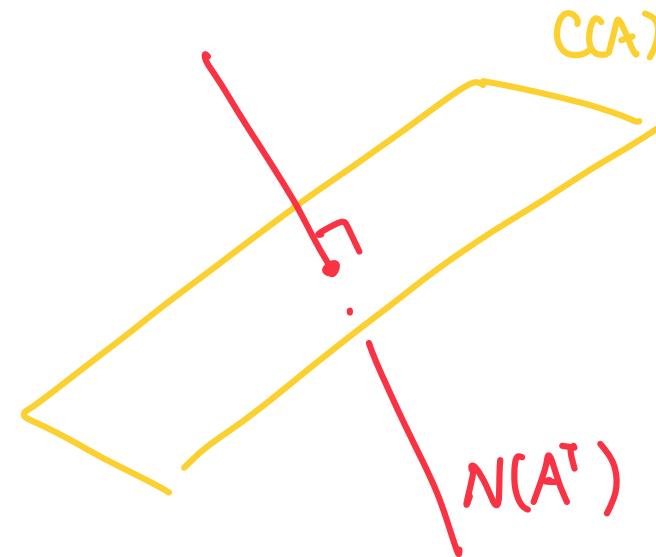


- L_A send $N(A)$ to $\{\vec{0}\}$, L_{A^\top} send $N(A^\top)$ to $\{\vec{0}\}$.
- $L_A : C(A^\top) \longrightarrow C(A)$ is an isomorphism

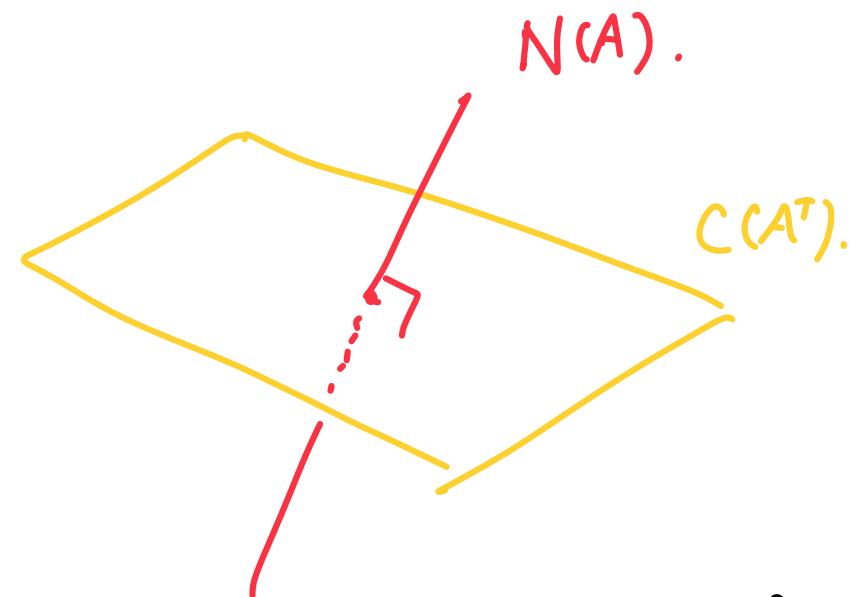
orthogonality complement

Picture:

\mathbb{R}^m :



\mathbb{R}^n :

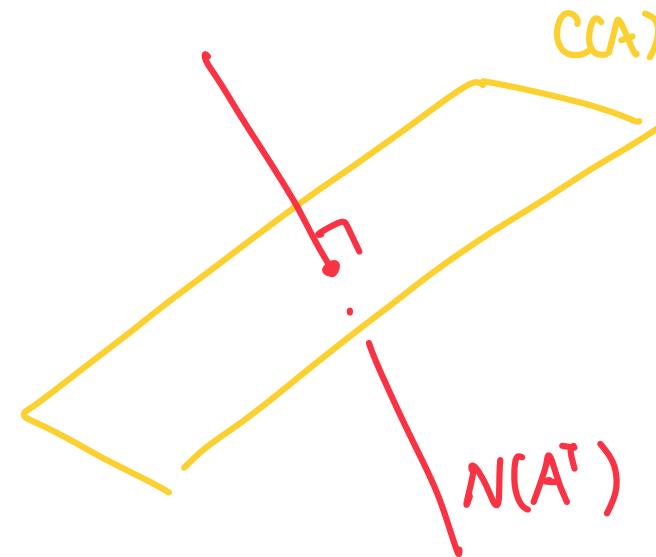


- L_A send $N(A)$ to $\{\vec{0}\}$, L_{A^\top} send $N(A^\top)$ to $\{\vec{0}\}$.
- $L_A : C(A^\top) \rightarrow C(A)$ is an isomorphism
 $\therefore \text{Ker}(L_A) = \{\vec{0}\}, \text{Im}(L_A) = C(A)$

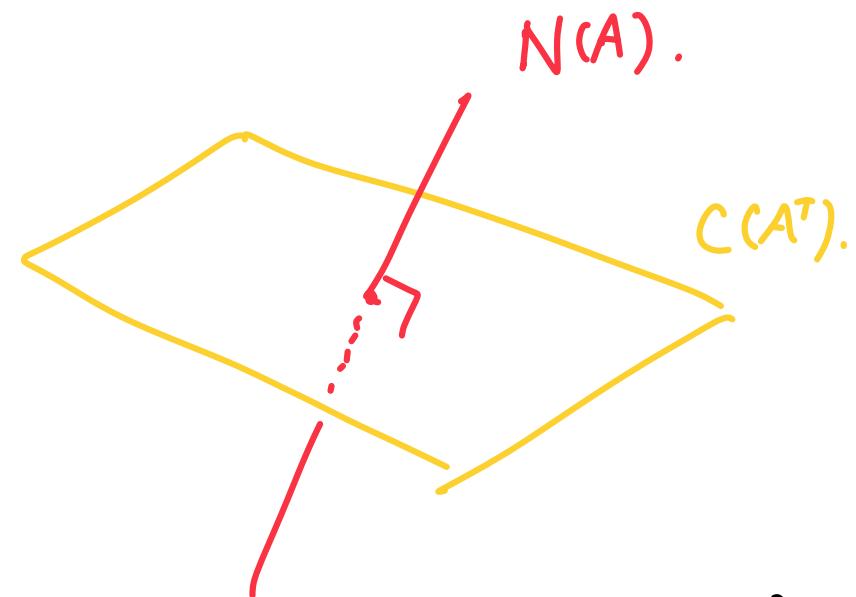
orthogonality complement

Picture:

\mathbb{R}^m :



\mathbb{R}^n :

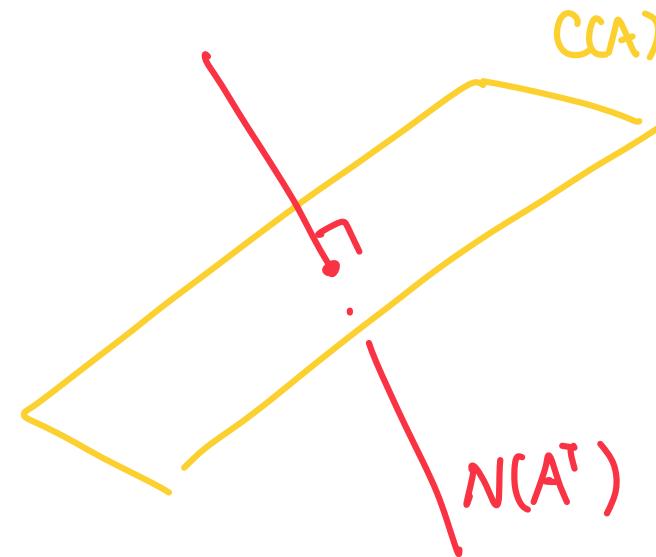


- L_A send $N(A)$ to $\{\vec{0}\}$, L_{A^\top} send $N(A^\top)$ to $\{\vec{0}\}$.
- $L_A : C(A^\top) \rightarrow C(A)$ is an isomorphism
 $\because \text{Ker}(L_A) = \{\vec{0}\}, \text{Im}(L_A) = C(A)$
- $L_{A^\top} : C(A) \rightarrow C(A^\top)$ iso.

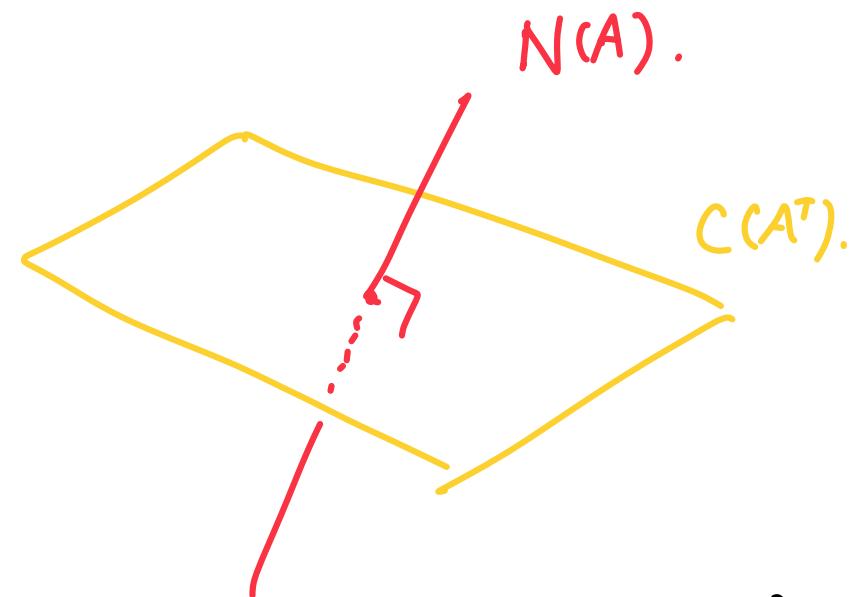
orthogonality complement

Picture:

\mathbb{R}^m :



\mathbb{R}^n :



- L_A send $N(A)$ to $\{\vec{0}\}$, L_{A^T} send $N(A^T)$ to $\{\vec{0}\}$.
- $L_A : C(A^T) \rightarrow C(A)$ is an isomorphism
 $\because \text{Ker}(L_A) = \{\vec{0}\}, \text{Im}(L_A) = C(A)$
- $L_{A^T} : C(A) \rightarrow C(A^T)$ iso.

Warning:

$L_{A^T} \circ L_A = \text{id}$ they are not inverse.