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#### Definition

Inner Product: An inner product is a function  $\langle \cdot, \cdot \rangle : V \times V \to \mathbb{F}$  with the following properties for all  $\mathbf{u}, \mathbf{v}, \mathbf{w} \in V$ , and  $a, b \in \mathbb{F}$ .

1.  $\langle \mathbf{u}, \mathbf{v} \rangle = \overline{\langle \mathbf{v}, \mathbf{u} \rangle}$  Note: If  $\mathbb{F} = \mathbb{R}$ , then we omit the conjugate!

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Let  $V = \mathbb{C}^n$  and  $\mathbb{F} = \mathbb{C}$ . Then the following function is an inner product

$$\langle \mathbf{x}, \mathbf{y} \rangle = \overline{\mathbf{y}}^{\top} \mathbf{x} = \sum_{k=1}^{n} x_k \overline{y_k}$$

Note: we sometimes abbreviate  $\overline{\mathbf{v}}^{\top}$  as  $\mathbf{v}^*$ 

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Note that if we are in the real numbers, then we omit the conjugate of y.

Note: we sometimes abbreviate  $\overline{\mathbf{y}}^{\top}$  as  $\mathbf{y}^*$ 

Property 1:  $\langle \mathbf{x}, \mathbf{y} \rangle = \overline{\langle \mathbf{y}, \mathbf{x} \rangle}$ 

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In addition, the only time  $\sum_{k=1}^{n} |x_k|^2 = 0$  is when all components are 0 or if  $\mathbf{x} = \mathbf{0}$ 

#### Definition

Dot Product: The dot product of two vectors in  $\mathbb{R}^n$  is a function given by

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Note: For this course, we will only consider this inner product unless stated otherwise

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### Induced Norms

### **Theorem**

Let  $(V, \mathbb{F})$  be a vector space with some inner product  $\langle \cdot, \cdot \rangle$ . Then the induced norm of this space is

$$\|\mathbf{x}\| = \sqrt{\langle \mathbf{x}, \mathbf{x} \rangle}$$

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### Proof.

We can prove all 3 properties from the previous slide as consequences of us using the inner product.



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For this course, we will consider only this norm unless stated otherwise.

### **Unit Vector**

### Definition

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$$||x|| = 1$$

If we have any vector,  $\mathbf{x} \neq \mathbf{0}$ , then we can find a vector pointing in the same direction, but is also of *unit length* by doing:

$$\mathbf{y} = \frac{\mathbf{x}}{\|\mathbf{x}\|}$$

### **Distance**

For us, we can think of distance between vectors as "how large is the difference between two vectors", or in other words, we say

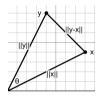
$$d(\mathbf{x}, \mathbf{y}) = \|\mathbf{y} - \mathbf{x}\|$$

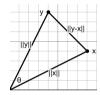
Note: this idea is closely related to "Metric Spaces" 1, which you will see in an analysis course.

Johnathan Rhyne (CU Denver) Math 3191 Inner Products and Orthogonality

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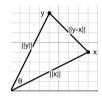
<sup>1</sup>https://en.wikipedia.org/wiki/Metric\_space#Definition





Using the law of cosines<sup>a</sup>, we have that

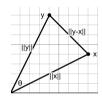
$$\left\|\mathbf{y}-\mathbf{x}\right\|^{2}=\left\|\mathbf{x}\right\|^{2}+\left\|\mathbf{y}\right\|^{2}-2\left\|\mathbf{x}\right\|\left\|\mathbf{y}\right\|\cos(\theta)$$



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$$\|\mathbf{y} - \mathbf{x}\|^2 = \|\mathbf{x}\|^2 + \|\mathbf{y}\|^2 - 2\|\mathbf{x}\| \|\mathbf{y}\| \cos(\theta)$$

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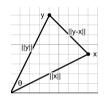
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Which (assuming that  $\mathbf{x}, \mathbf{y} \neq \mathbf{0}$ ) can be solved for  $\theta$  to get

$$\theta = \cos^{-1}\left(\frac{\mathbf{x}^{\top}\mathbf{y}}{\|\mathbf{x}\|\,\|\mathbf{y}\|}\right)$$

<sup>a</sup>https:

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In higher dimensions and other vector spaces, this is how we define the angle between vectors

//en.wikipedia.org/wiki/Law\_of\_cosines

<sup>&</sup>lt;sup>a</sup>https:

# Orthogonality

Let  $(V, \mathbb{F})$  be a vector space where V denotes the set our vectors come from,  $\mathbb{F}$  is the set our scalars come from, and we have some inner product  $\langle \cdot, \cdot \rangle$ .

### Definition

Orthogonal Vectors: We say that 2 vectors,  $(\mathbf{x}, \mathbf{y} \in V)$  are orthogonal (or perpendicular) if

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Note that assuming we have non-zero vectors, the angle between x, y would be 90°!

Since orthogonality is closely tied to our inner product, we will use our standard one for this course.

# Special Case for Orthogonality

#### **Theorem**

If  $V = \mathbb{R}^n$  (or equivalently  $\mathbb{C}^n$ ) with the usual inner product, then  $\mathbf{0}$  is orthogonal to every vector.

#### Proof.

Let  $\mathbf{x} \in \mathbb{R}^n$  and  $\langle \cdot, \cdot \rangle$  denote the standard inner product, then we have that

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### Proof.

Let  $\mathbf{x} \in \mathbb{R}^n$  and  $\langle \cdot, \cdot \rangle$  denote the standard inner product, then we have that

$$\langle \mathbf{0}, \mathbf{x} \rangle = \sum_{k=1}^{n} 0 \cdot x_k = 0$$