Geom	etric Si	gnal T	heory
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# Geometric Signal Theory

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Vector Notation

#### Spaces

### **Geometric Signal Theory**

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## • Inner Product Space

**Inner Product:** 

$$\langle x \ y \rangle \stackrel{\Delta}{=} \sum_{n=0}^{N-1} x(n) \overline{y(n)}$$
 where  $x, y \in \mathbf{C}^N$ 

some Inner Product properties:

1: 
$$\langle x - y, z \rangle = \langle x, z \rangle - \langle y, z \rangle$$

$$2: \langle \alpha x, y \rangle = \alpha \langle x, y \rangle$$

$$3: \langle x, \alpha y \rangle = \overline{\alpha} \langle x, y \rangle$$

Inner Product is linear.

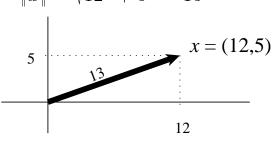
Norm:

$$||x||^2 \triangleq \langle x | x \rangle$$

example:

$$||x|| = \sqrt{12^2 + 5^2} = 13$$

The Norm of a vector is its "length".



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**Properties of the Norm:** 

- $1: ||x|| = 0 \iff x = 0$
- $2: \|x + y\| \le \|x\| + \|y\|$
- 3: for  $c \in \mathbb{C}$ ,  $||cx|| = |c| \cdot ||x||$
- 4:  $||x|| \cdot ||y|| \ge |\langle x, y \rangle|$  (Schwartz inequality)

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Orthogonality

$$x \perp y \iff \langle x, y \rangle = 0$$

#### **Geometric Signal Theory**

#### Projection

orthogonal projection:

$$\mathcal{P}_{x}(y) = x_{y} \stackrel{\Delta}{=} \left(\frac{\langle y, x \rangle}{\|x\|^{2}}\right) x$$

example:

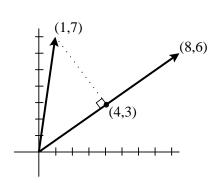
$$\mathcal{P}_{x}(y) = \frac{\langle y, x \rangle}{\|x\|^{2}} x$$

$$= \frac{\langle (1,7), (8,6) \rangle}{\|(8,6)\|^{2}} (8,6)$$

$$= \frac{8+42}{64+36} (8,6)$$

$$= \frac{1}{2} (8,6)$$

$$= (4,3)$$



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### Signal Measures

Mean (average):

$$\mu_x \triangleq \frac{1}{N} \sum_{n=0}^{N-1} x_n$$

$$x \in \mathbf{C}^N$$

**Energy:** 

$$\mathcal{E}_x \triangleq \langle x | x \rangle = ||x||^2 = \sum_{n=0}^{N-1} |x|^2$$

Power:

$$P_x \triangleq \frac{1}{N} \mathcal{E}_x = \frac{1}{N} \sum_{n=0}^{N-1} |x_n|^2$$

- also called the mean square
- represents the average energy per sample

#### Variance:

$$\sigma_x^2 \triangleq P_x - \mu_x^2$$

"mean square" - "squared mean"

$$\sigma_x^2 = \frac{1}{N} \sum_{n=0}^{N-1} (x_n - \mu_x)^2$$

- Removes "DC" component of signal - gives a better feel for the signal level.

#### **Geometric Signal Theory**

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#### Basis Vectors

A set of mutually orthogonal vectors

"Natural" Basis (Catesian Basis)

- Basis set for samples in a signal

$$\hat{e}_0 = [1, 0, 0]$$

$$\hat{e}_0 = [1, 0, 0]$$
  $\hat{e}_1 = [0, 1, 0]$   $\hat{e}_2 = [0, 0, 1]$ 

$$\hat{e}_2 = [0, 0, 1]$$

"Sinusoidal" Basis

- Basis set for samples in a spectrum

$$\hat{s}_k(n) \triangleq e^{j\omega_k nT} = e^{j2\pi kn/N}$$

3D-example:

$$\begin{split} \hat{s}_0 &= \left. e^{j2\pi0n} \right|_{n=0}^3 = [1,\,1,\,1] \\ \hat{s}_1 &= \left. e^{j2\pi1n} \right|_{n=0}^3 = [1,\,e^{j2\pi/3},\,e^{j4\pi/3}] = [1,\,\frac{-1+j\sqrt{3}}{2},\,\frac{-1-j\sqrt{3}}{2}] \\ \hat{s}_2 &= \left. e^{j2\pi2n} \right|_{n=0}^3 = [1,\,e^{j4\pi/3},\,e^{j8\pi/3}] = [1,\,\frac{-1-j\sqrt{3}}{2},\,\frac{-1+j\sqrt{3}}{2}] \end{split}$$

$$\hat{s}_0 \perp \hat{s}_1 \perp \hat{s}_2$$

#### **Orthonormal Basis**

The natural basis vectors all have a norm of 1. However, the sinusoidal basis vectors all have a norm of N.

#### **Orthonormal Sinusoidal Basis:**

$$\tilde{s}_k \triangleq \frac{\hat{s}_k}{\|\hat{s}_k\|}$$