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S3-Euclidean Spaces PART I

Linear Algebra

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Plan



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- Euclidean Spaces. Definition
- Dot products in \mathbb{R}^2 , \mathbb{R}^3 and \mathbb{R}^n .
- Matrix form of scalar products: the Gram matrix
- Length and angle. Orthogonality.
- Euclidean spaces of digital signals, analogic signals and images.
- Orthogonal Projections.
- Gram-Schmidt method.
- To practice



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<https://www.khanacademy.org/math/linear-algebra/vectors-and-spaces/dot-cross-products/v/vector-dot-product-and-vector-length>

Euclidean Spaces



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If V is a vector space, a dot product (or scalar product) is any application \cdot :

$$\begin{aligned} \cdot : V \times V &\rightarrow \mathbb{R} \\ (\vec{v}, \vec{u}) &\rightarrow \vec{v} \cdot \vec{u} \end{aligned}$$

fulfilling the following conditions:

1. Linearity

$$\forall \vec{v}_1, \vec{v}_2, \vec{u} \in V \quad \forall \alpha, \beta \in \mathbb{R}$$

$$\begin{aligned} (\alpha \vec{v}_1 + \beta \vec{v}_2) \cdot \vec{u} &= \alpha \vec{v}_1 \cdot \vec{u} + \beta \vec{v}_2 \cdot \vec{u} \\ \vec{u} \cdot (\alpha \vec{v}_1 + \beta \vec{v}_2) &= \alpha \vec{u} \cdot \vec{v}_1 + \beta \vec{u} \cdot \vec{v}_2 \end{aligned}$$

2. Symmetry

$$\forall \vec{v}, \vec{u} \in V \quad \vec{v} \cdot \vec{u} = \vec{u} \cdot \vec{v}$$

3. Positivity

$$\forall \vec{v} \in V \quad \vec{v} \cdot \vec{v} \geq 0 \quad \vec{v} \cdot \vec{v} = 0 \rightarrow \vec{v} = \vec{0}$$



Euclidean Spaces

- The Vector space **V** provided with the scalar product \cdot , (V, \cdot) is called **Euclidean Space**
- The scalar products serve to **compare vectors**, for instance, to compare two digital or analogic signals and see if they are similar or not.

Norm (or length)of a vector: $\|\vec{v}\| = \sqrt{\vec{v} \cdot \vec{v}} = \sqrt{\vec{v}^2}$.

The norm of a vector is its length according to the the scalar product \cdot .

Note: $\frac{\vec{v}}{\|\vec{v}\|}$ is the unit vector in the direction of \vec{v} .

Scalar product in \mathbb{R}^n

- **Euclidean scalar product**

$$\mathbf{u} \cdot \mathbf{v} = \sum_{k=1}^n u_k \cdot v_k = u_1 v_1 + u_2 v_2 + \cdots + u_n v_n$$

satisfies these properties of scalar product.

Prove the 3 conditions that has to fulfill any scalar product.

- **The norm:**

$$\|\mathbf{u}\| = \sqrt{\mathbf{u} \cdot \mathbf{u}} = \sqrt{u_1^2 + \cdots + u_n^2} > 0.$$



Scalar product in Matrix form

- $B = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ basis set of V
- $\mathbf{x} = x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \dots + x_n \mathbf{v}_n$
- $\mathbf{y} = y_1 \mathbf{v}_1 + y_2 \mathbf{v}_2 + \dots + y_n \mathbf{v}_n$
- $\mathbf{x} \cdot \mathbf{y} = (x_1 \mathbf{v}_1 + x_2 \mathbf{v}_2 + \dots + x_n \mathbf{v}_n) \cdot (y_1 \mathbf{v}_1 + y_2 \mathbf{v}_2 + \dots + y_n \mathbf{v}_n)$

$$\mathbf{x} \cdot \mathbf{y} = (x_1, x_2, \dots, x_n) \cdot \mathbf{G} \cdot \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \text{ with } \mathbf{G} = \begin{pmatrix} \mathbf{v}_1 \cdot \mathbf{v}_1 & \cdots & \mathbf{v}_1 \cdot \mathbf{v}_n \\ \vdots & \ddots & \vdots \\ \mathbf{v}_n \cdot \mathbf{v}_1 & \cdots & \mathbf{v}_n \cdot \mathbf{v}_n \end{pmatrix}$$

$$\mathbf{x} \cdot \mathbf{y} = \mathbf{x}^T \cdot \mathbf{G} \cdot \mathbf{y}, \text{ with } \mathbf{y} = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} \text{ and } \mathbf{x}^T = (x_1, x_2, \dots, x_n).$$

\mathbf{G} is called the Gram matrix of the scalar product in the basis $B = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$.

Note : \mathbf{x}^T is always row vector !!



Gram matrix

The Gram matrix has some important properties that come from the properties of the scalar properties

- Symmetry: $G(i, j) = \mathbf{v}_i \cdot \mathbf{v}_j = G(j, i) = \mathbf{v}_j \cdot \mathbf{v}_i \leftrightarrow G = G^T$
- Definite positive $\mathbf{x}^T \cdot G \cdot \mathbf{x} > 0 \quad \forall \mathbf{x} \neq \mathbf{0}$

Note: any matrix G fulfilling these two conditions correspond to scalar product in V

$$G = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \quad \mathbf{x}^T \cdot G \cdot \mathbf{x} = (x_1, x_2) \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 2x_1^2 + 2x_1x_2 + x_2^2 = x_1^2 + (x_1 + x_2)^2 > 0.$$



What happens with the Gram matrix when we change the basis?

- $B = \{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ basis set of V
- $B' = \{\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n\}$ basis set of V

$$\bullet \mathbf{x} \cdot \mathbf{y} = (x_1, x_2, \dots, x_n)_B \cdot \mathbf{G} \cdot \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix}_B = (c_1, c_2, \dots, c_n)_{B'} \cdot \mathbf{G}^* \cdot \begin{pmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{pmatrix}_{B'}$$

- $\mathbf{x} = \mathbf{P} \cdot \mathbf{c}$; $\mathbf{y} = \mathbf{P} \cdot \mathbf{s} \rightarrow \mathbf{x} \cdot \mathbf{y} = (\mathbf{P} \mathbf{c})^T \cdot \mathbf{G} \cdot (\mathbf{P} \mathbf{s})$
- $\mathbf{G}^* = \mathbf{P}^T \mathbf{G} \mathbf{P}$ (Congruent matrices)



Bit of Practice

❖ In \mathbb{R}^2 we define the following dot product

$$(x_1, x_2) \cdot (y_1, y_2) = 2x_1y_1 + x_2y_1 + x_1y_2 + x_2y_2$$

- ✓ Finding the Gram matrix in the canonic basis.
- ✓ Finding the Gram matrix in the *basis* $B = \{(1, 1), (-1, 1)\}$
- ✓ Finding the norm of $(1, 1)$.



Angle between 2 vectors

Angle that form 2 vectors

$$\mathbf{u} \cdot \mathbf{v} = \|\mathbf{u}\| \|\mathbf{v}\| \cos\langle \mathbf{u}, \mathbf{v} \rangle$$

$$-1 < \cos\langle \mathbf{u}, \mathbf{v} \rangle = \frac{\mathbf{u} \cdot \mathbf{v}}{\|\mathbf{u}\| \|\mathbf{v}\|} < 1 \quad (\text{Cauchy-Schwartz inequality})$$

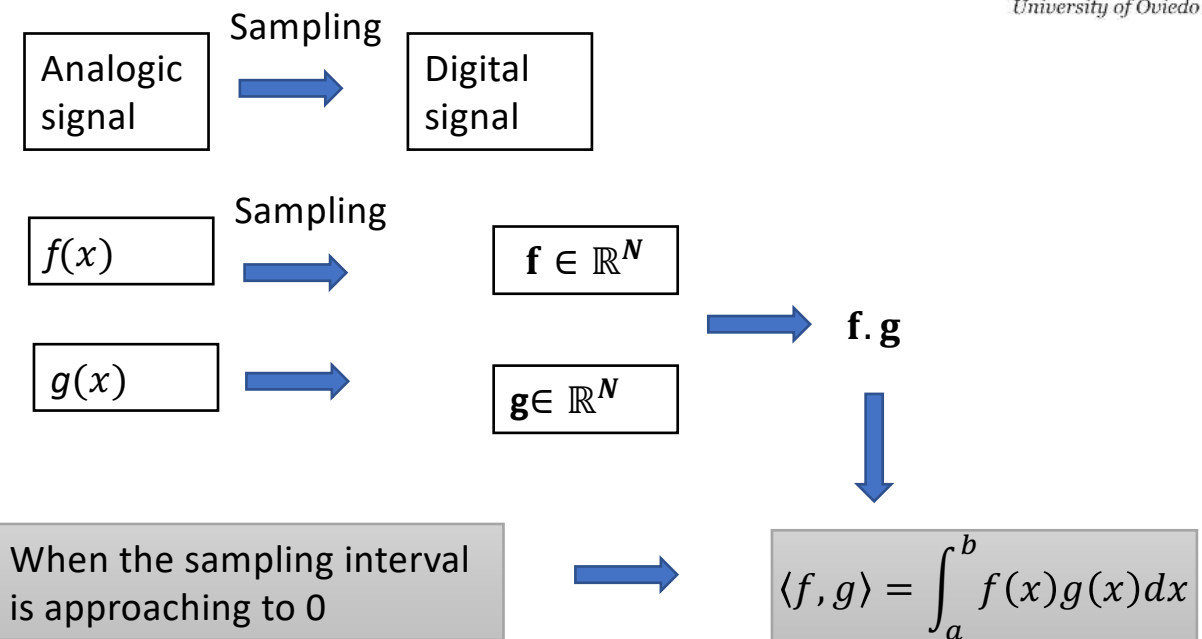
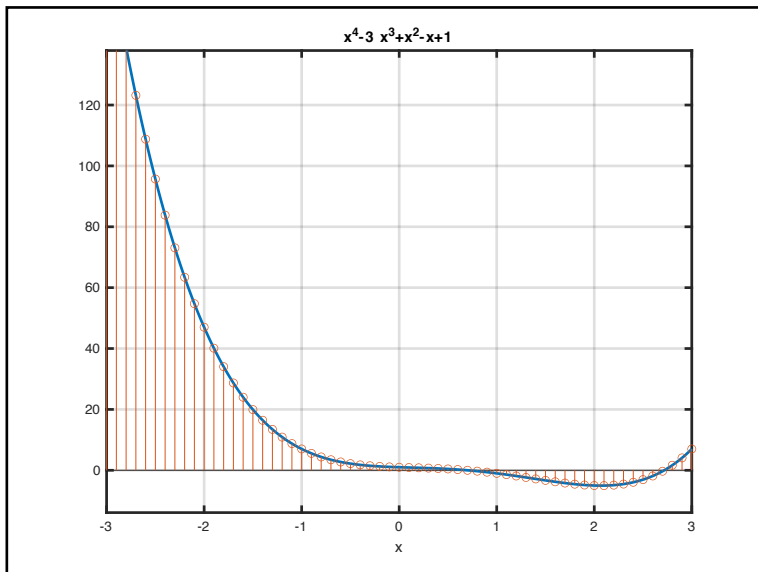
$$-\|\mathbf{u}\| \|\mathbf{v}\| < \mathbf{u} \cdot \mathbf{v} < \|\mathbf{u}\| \|\mathbf{v}\|$$

$$(\mathbf{u} + \mathbf{v})^2 = (\mathbf{u} + \mathbf{v}) \cdot (\mathbf{u} + \mathbf{v}) = \|\mathbf{u} + \mathbf{v}\|^2 = \|\mathbf{u}\|^2 + \|\mathbf{v}\|^2 + 2\mathbf{u} \cdot \mathbf{v} \leq (\|\mathbf{u}\| + \|\mathbf{v}\|)^2$$

$$\|\mathbf{u} + \mathbf{v}\| \leq \|\mathbf{u}\| + \|\mathbf{v}\| \quad (\text{Triangle Inequality})$$



Scalar product in $C^0_{[a,b]}$



$$\cos \langle f, g \rangle = \frac{f \cdot g}{\|f\| \|g\|} = \frac{\int_a^b f(x)g(x)dx}{\sqrt{\int_a^b f^2(x)dx} \sqrt{\int_a^b g^2(x)dx}}$$

Bit of Practice



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Considering in $C_{[0,1]}^0$ 3 functions:

$$f(x) = e^x, \quad g(x) = x, \quad h(x) = x^2,$$

1. Finding the norms of these functions
2. Deciding which pairs of functions are more similar: (f, g) , (f, h) , (h, g) .



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Entrywise Scalar Product in $M_{m \times n}(\mathbb{R})$



- We saw that gray color images are matrices of integers.
- The question is how to compare images?
- $A, B \in M_{m \times n}(\mathbb{R})$
- The simplest way is: $A \cdot B = A(:,)^T G B(:,)$ **(elementwise scalar product)**

Comparing Matrices

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{pmatrix} \quad A \cdot B^T = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{pmatrix} \begin{pmatrix} b_{11} & b_{21} \\ b_{12} & b_{22} \\ b_{13} & b_{23} \end{pmatrix} =$$

$$B = \begin{pmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{pmatrix} = \begin{pmatrix} a_{11}b_{11} + a_{12}b_{12} + a_{13}b_{13} & B \\ C & a_{21}b_{21} + a_{22}b_{22} + a_{23}b_{23} \end{pmatrix}$$

$$\boxed{\text{trace}(AB^T) = \text{trace}(BA^T) = \vec{A}(:,)^T \cdot \vec{B}(:,)}$$