

# MIMF Lecture Week 2, Day 3: Orthonormality, Types of Matrices, Linear Functionals, Riesz Representation Theorem

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## Orthonormality

**Definition:** A set of vectors  $\{\mathbf{u}_1, \dots, \mathbf{u}_m\} \subset \mathbb{R}^n$  is *orthonormal* if

$$\langle \mathbf{u}_i, \mathbf{u}_j \rangle = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases}$$

**Example:** Check whether  $\{(1, 0, 1), (1, 1, 0), (0, 1, 1)\}$  is orthonormal.

$$\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = \underline{\hspace{2cm}}.$$

$$\langle \mathbf{v}_1, \mathbf{v}_3 \rangle = \underline{\hspace{2cm}}.$$

$$\|\mathbf{v}_1\| = \underline{\hspace{2cm}}.$$

So the set (is / is not) orthonormal.

## Applications:

- **Quant finance:** Principal component analysis (PCA) of covariance matrices uses orthonormal eigenvectors to identify uncorrelated risk factors.

- **Physics:** Quantum states form orthonormal bases in Hilbert spaces.
- **CS / ML:** Feature orthogonalization improves model interpretability.

## Types of Matrices

### Definitions:

- A matrix  $A \in M_{m \times n}(\mathbb{R})$  is *symmetric* if \_\_\_\_\_.
- A matrix  $A \in M_{m \times n}(\mathbb{F})$  is *Hermitian/self-adjoint* if \_\_\_\_\_.
- A matrix  $A \in M_{m \times n}(\mathbb{R})$  is *orthogonal* if \_\_\_\_\_.
- A matrix  $U \in M_{m \times n}(\mathbb{F})$  is *unitary* if \_\_\_\_\_.
- A matrix  $U \in M_{m \times n}(\mathbb{F})$  is *normal* if \_\_\_\_\_.
- A matrix  $A \in M_{m \times n}(\mathbb{F})$  is *positive semi-definite* if:
  - 1) \_\_\_\_\_ for all  $\mathbf{x} \in \mathbb{F}^n$ ; and,
  - 2)  $A$  is Hermitian/self-adjoint.

What we care about: A positive semi-definite matrix has all \_\_\_\_\_ eigenvalues.

- A matrix  $A \in M_{m \times n}(\mathbb{F})$  is *positive definite* if:
  - 1) \_\_\_\_\_ for all  $\mathbf{x} \in \mathbb{F}^n$ ; and,
  - 2)  $A$  is Hermitian/self-adjoint.

A positive definite matrix has all \_\_\_\_\_ eigenvalues.

### Examples:

$$A = \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix}, \quad B = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}, \quad C = \begin{bmatrix} 1 & 2 \\ 2 & 4 \end{bmatrix}$$

- $A$  is Hermitian: \_\_\_\_\_

- $A$  has non-negative eigenvalues: \_\_\_\_\_
- $A$  is Positive semi-definite: \_\_\_\_\_
- $B$  is hermitian: \_\_\_\_\_
- $B$  is Positive semi-definite: \_\_\_\_\_
- $C$ : Left as an exercise to the reader :)

### Applications:

- **Quant finance:** Covariance matrices (symmetric, PSD) underpin portfolio risk modeling.
- **Physics:** Hermitian operators correspond to observables with real eigenvalues.
- **CS:** Orthogonal/unitary matrices are used in QR decomposition and FFTs.

## Linear Functionals and Riesz Representation

**Definition:** A *linear functional* is a linear map  $L : \text{_____} \rightarrow \text{_____}$

In English: A linear functional takes in something from a vector space and outputs a scalar in its underlying field.

### Applications:

- **Quant finance:** Portfolio valuation and risk measures are linear functionals of asset payoffs. They take in vectors of information and output scalar risk measures.
- **Physics:** Fourier expansions use Riesz to represent functionals.
- **CS:** Duality in optimization and kernel methods rely on linear functionals.

**Riesz Representation Theorem (RRT):** Every linear functional  $L$  on  $\mathbb{F}^n$  can be written as

$$L(\mathbf{x}) = \underline{\hspace{2cm}} \quad \text{for a unique } \mathbf{y} \in \mathbb{F}^n.$$

In english: there exists a vector  $\mathbf{y} \in \mathbb{F}^n$  such that the linear functional on  $\mathbf{x}$ ,  $L(\mathbf{x})$ , can be written as the inner product of  $\mathbf{x}$  with  $\mathbf{y}$ .

**Example:**

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}; \quad L(\mathbf{x}) = 2x_1 - 3x_2 + 5x_3$$

$$\text{Answer: } \mathbf{y} = \underline{\hspace{3cm}}$$

Test:

$$\mathbf{x} = \begin{bmatrix} 4 \\ 1 \\ 7 \end{bmatrix} \quad L(\mathbf{x}) = \underline{\hspace{3cm}}.$$

$$\mathbf{y} = \begin{bmatrix} 2 \\ -3 \\ 5 \end{bmatrix} \quad \langle \mathbf{x}, \mathbf{y} \rangle = \underline{\hspace{3cm}}.$$

What vector space does  $L$  pull from?  $\underline{\hspace{2cm}}$