

ECE 269: Linear Algebra and Applications

Lecture 1: Introduction

Behrouz Touri (email: btouri@ucsd.edu, office: EBU1 6408)

University of California San Diego

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Course Logistics

This course: Teaching crew :-)

- Behrouz Touri (instructor)
 - Office: EBU1 (JH) 6408
 - Email: btouri@ucsd.edu
 - Office hours: Tuesdays 11:00 am to 12:00 pm (in-person)
 - Research interest: control theory, dynamics over networks, stochastic systems

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- Zhuqing Li (TA)
 - Email: zh1160@ucsd.edu
 - Office hours: Weds 11:00 am to 12:00 pm (through Zoom (link is available on Canvas))

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 - Email: zhl160@ucsd.edu
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 - Office: TBD
 - Email: miz152@ucsd.edu
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- Anand Kumar (TA)
 - Office: Jacobs Hall (EBU1) 5101C
 - Email: ank029@ucsd.edu
 - Office hours: Thursdays 11:00 am to 12:00 pm (in-person)

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- Example: copying Problem 1 in HW 1 results in 0 out of 15% of HW grade for the course.

Intro to Linear Algebra

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- Example topics:
 - (i) What scalar, addition, scalar multiplication means?
 - (ii) Moving beyond Euclidean spaces
 - (iii) What vectors can be achieved by linear transformations?
 - (iv) What happens when we combine linear transformations?
 - (v) ...

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- Optimization theory: pretty much anything! :-)

Real-world problem:

- Given the movie ratings of a number of women and men and knowing their genders, can we determine the gender of users with unspecified gender based on their preferences?

Applications of Linear Algebra: Machine Learning

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A Mathematical approach to this problem:

- Suppose we are given two sets of vectors $A = \{x_1, \dots, x_n\}$ and $B = \{y_1, \dots, y_m\}$ in \mathbb{R}^n , representing the ratings of the users in each group, where n is the selected number of movies.

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- Can we find a linear subspace separating the two sets?

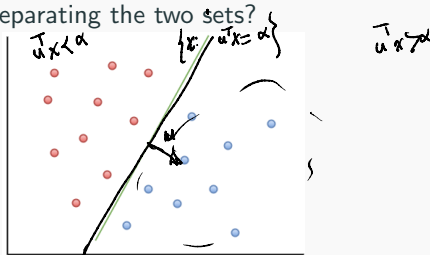
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Applications of Linear Algebra: Control Theory

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- Suppose that we have an imperfect communication channel. How to send information over this channel *reliably* by introducing redundant information.

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- Suppose that u is a binary vector of length k representing the information to be transformed.
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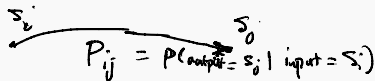
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- Suppose that we are given a non-negative matrix P such that rows and columns of P add to 1.

Applications of Linear Algebra: Markov Chains



Handwritten diagram showing a transition from state s_i to state s_j with the probability $P_{ij} = P(\text{output} = s_j \mid \text{input} = s_i)$.

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- Suppose that we are given a non-negative matrix P such that rows and columns of P add to 1.
- The above question is very closely related to the following problem:
Suppose that $P^k \rightarrow A$ as $k \rightarrow \infty$. Then, what can we say about the speed of convergence.

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$$\mathbb{Z}_2 = \{0, 1\}$$

$$p \geq 2$$

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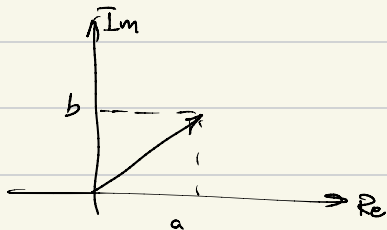
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 - d. \mathbb{C} := $\{(a + bj) \mid a, b \in \mathbb{R}\}$ the set of complex numbers

$$j = \sqrt{-1}$$

$$\text{set of complex numbers} = \mathbb{C} = \{a+bj \mid a, b \in \mathbb{R}\} \equiv \{(a, b) \mid a, b \in \mathbb{R}\}$$

$$c = (a+bj) : \operatorname{Re}(c) = a$$

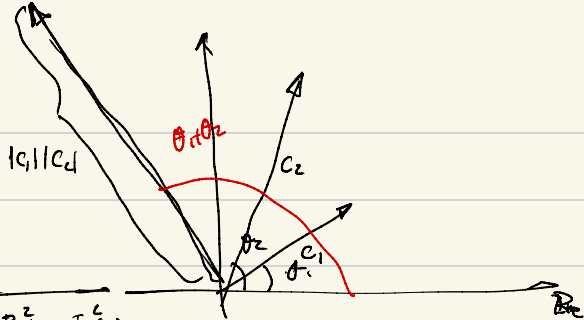
$$\operatorname{Im}(c) = b$$



$$c_1 = a_1 + b_1 j, c_2 = a_2 + b_2 j \Rightarrow c_1 + c_2 = (a_1 + a_2) + (b_1 + b_2) j$$

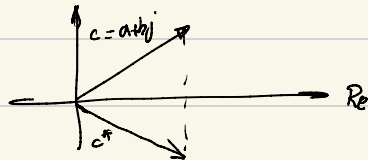
$$c_1 c_2 = (a_1 a_2 - b_1 b_2) + (a_1 b_2 + a_2 b_1) j$$

$$(a_1 + b_1 j)(a_2 + b_2 j) = a_1 a_2 + a_1 b_2 j + a_2 b_1 j + b_1 b_2 \underbrace{j^2}_{-1} = (a_1 a_2 - b_1 b_2) + (a_1 b_2 + a_2 b_1) j$$



notation: $|c| = \sqrt{\text{Re}(c)^2 + \text{Im}(c)^2}$

for $c = a + bj$, we define $c^* = \bar{c} := a - bj$



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 - d. $\mathbb{C} := \{(a + bj) \mid a, b \in \mathbb{R}\}$ the set of complex numbers
- We use capital letters A, B, Γ , etc., to denote matrices (most often).
- For a set \mathcal{X} , the set of $m \times n$ matrices with entries in \mathcal{X} is denoted by $\mathcal{X}^{m \times n}$.
- We use capital letters A, B, Γ , etc., to denote matrices (most often)
- We use A_{ij} to denote the (i, j) th entry of matrix A .

Basic Notations

- In this course, all vectors are considered to be column vectors.
- Notations:
 - a. \mathbb{Z} : the set of integers $\dots, -2, -1, 0, 1, 2, \dots$
 - b. $\mathbb{Z}_p : \{0, 1, \dots, p - 1\}$
 - c. \mathbb{R} : the set of real numbers
 - d. $\mathbb{C} := \{(a + bj) \mid a, b \in \mathbb{R}\}$ the set of complex numbers
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- We use capital letters A, B, Γ , etc., to denote matrices (most often)
- We use A_{ij} to denote the (i, j) th entry of matrix A .
- For a matrix A , we denote its transpose by A' or A^T .

Fields: A field $(F, +, \cdot)$ such that
set \downarrow
 $\cdot, +: F \times F \rightarrow F$

$+$ needs to:

$$(i): a+b = b+a$$

$$(ii): \underset{\substack{\uparrow \\ \text{there exists}}}{\exists} z \in F : a+z = z+a = a \quad \forall a \in F \quad \underset{\substack{\uparrow \\ \text{for all}}}{\forall}$$

$$(iii): \forall a \in F, \exists! b \in F : a+b = b+a = z$$

\downarrow
there exists a unique element

$$(iv): (a+b)+c = a+(b+c) \quad \forall a, b, c \in F$$

i.e., $(F, +)$ is an Abelian (Commutative) group

(F, \cdot) is also an Abelian group

\downarrow
 $F - \{z\}$

\downarrow
identity element of $+$