

LECTURE 1 SUMMARY

- Syllabus
- COURSE SUMMARY:

WHAT: Solving $Ax=b$
Spectral decompositions
WHY: SCIENCE, ENGINEERING, DATA
- NOTATION

• Scalar quantities ($\mathbb{R}, \mathbb{C}, \mathbb{N}$) ^{real} ^{complex} ^{integer} greek lower case
exception: $m, n, i, j, k, \ell \in \mathbb{N}$

• Vectors: lower case roman ($x, y, w, u \dots$)

All vectors will be column vectors.

v^* will indicate

α row vector whose elements are the complex conjugate of v .

$$v = \begin{bmatrix} v(1) \\ v(2) \\ v(3) \end{bmatrix} ; v^* = \begin{bmatrix} \overline{v(1)} & \overline{v(2)} & \overline{v(3)} \end{bmatrix} ;$$

$v(1) = a + ib$
 $\overline{v(1)} = a - ib$

complex conjugate

$v(i)$ indicates the i th element of the vector.

Occasionally, to denote a set of vectors, I will use an underline. e.g. $\underline{v_1}, \underline{v_2}$ indicate vectors.

Matrices: uppercase roman: A, B, C, \dots
 & greek: Λ, Ξ

ABSTRACT VECTOR spaces, $\mathcal{V}, \mathcal{W}, \mathcal{X}$

Addition: Associativity, Commutativity, Identity, Negative

Multiplication with scalar: Distributivity, Identity, Associativity

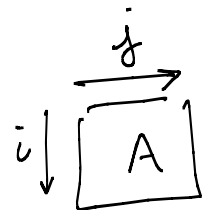
Matrices as a set of column vectors

e.g.

$$A = \begin{bmatrix} | & | & | & | \\ a_1 & a_2 & a_3 & a_4 \\ | & | & | & | \end{bmatrix}$$

Matrices as row vectors

$$A = \begin{bmatrix} \text{---} a_1^* \text{---} \\ \text{---} a_2^* \text{---} \\ \vdots \\ \text{---} a_n^* \text{---} \end{bmatrix}$$



Matrix entries

• elements

A_{ij} , $A(i, j)$

↑ indexes rows
↑ indexes columns

• columns

$A_{:i}$; $A(:, i)$

• rows

$A_{i:}$; $A(i, :)$

• Linear combinations of vectors

$\{v_1, \dots, v_n\}$; n vectors ; $v_i \in \mathbb{C}^m$ (or \mathbb{R}^m)

$\{\lambda_1, \dots, \lambda_n\}$ in scalars ; $\lambda_i \in \mathbb{C}$ (or \mathbb{R})

$w = \sum_{i=1}^n \lambda_i v_i$ is a linear combination of $\{v_i\}$

• SPAN of $\{v_i\}_{i=1}^n$; $v_i \in \mathbb{C}^m$
all possible linear combinations of $\{v_i\}$

• LINEARLY INDEPENDENT SETS OF VECTORS : $\nexists \{\lambda_i\}_{i=1}^n : \sum_i |\lambda_i| > 0$ AND $\sum_i \lambda_i v_i = 0$
THEN $\{v_i\}$ ARE LIN-IND.

• LINEARLY DEPENDENT SETS OF VECTORS : WE SAY $\{v_i\}_{i=1}^n$ ARE LIN-DEP IFF
 $\exists \{\lambda_i\}_{i=1}^n : \sum_i |\lambda_i| > 0$ AND $\sum_i \lambda_i v_i = 0$; $\lambda_i \in \mathbb{C}$; $v_i \in \mathbb{C}^n$.

• Basis of a vector space
Let V be a vector space. Let $\{v_i\}_{i=1}^n \in V$. It is a basis if
(1) $V = \text{SPAN} \{v_i\}$
(2) $\{v_i\}$ is lin-indep.

• Dimension of a vector space
 $\dim(V) = \#$ OF VECTORS IN A BASIS.

• Subspace of a vector space

• Inner product $x, y \in \mathbb{C}^n$
 $(x \cdot y) = x^* y = \sum_{j=1}^n \overline{x_j} y_j$

PROPERTIES

- $(x \cdot y) = \overline{(y \cdot x)}$
- $\alpha(x \cdot y) = (\alpha x) \cdot y$
- $(x+y, z) = (x, z) + (y, z)$
- $(x, x) > 0, \forall x \neq 0$

- Euclidean norm of a vector

$$\|x\|_2 = \sqrt{x \cdot x}$$

- Angle Between two vectors

$$\cos \alpha = (x \cdot y) / (\|x\|_2 \cdot \|y\|_2)$$

functions (or maps) of vectors

Linear maps

Let $f: V \rightarrow W$

(that is $v \in V; f(v) \in W$)

f is linear iff (if and only if)

$$f(\lambda_1 v_1 + \lambda_2 v_2) =$$

$$\lambda_1 f(v_1) + \lambda_2 f(v_2)$$

$$\forall \lambda_1, \lambda_2 \in \mathbb{C}; v_1, v_2 \in V$$

Given a linear map $f: \mathbb{C}^n \rightarrow \mathbb{C}^m$,

we can represent it as a matrix

$$Ax = \begin{bmatrix} | & & | \\ a_1 & \dots & a_n \\ | & & | \\ & & \vdots \\ & & | \\ & & x_n \end{bmatrix} = f(x)$$

$$= \sum x_i a_i, \text{ where } a_i = f(e_i)$$

and e_i is the canonical Basis in \mathbb{C}^n

$$e_1 = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}; e_2 = \begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}; \dots; e_n = \begin{bmatrix} 0 \\ \vdots \\ 1 \end{bmatrix}.$$

That is, applying a matrix to a vector gives another vector, which is a linear combination of the columns of the matrix.

ALTERNATIVELY, WE CAN VIEW A MATRIX VECTOR MULTIPLICATION AS A SET OF INNER PRODUCTS. GIVEN $a_i \in \mathbb{C}^n$ and $x \in \mathbb{C}^n$ WITH $i=1, \dots, m$ WE WISH TO COMPUTE $y_i = (a_i \cdot x)$ $i=1, \dots, m$

THIS OPERATION CAN BE WRITTEN AS

$$\begin{Bmatrix} y \\ y \\ y \\ \vdots \\ y \end{Bmatrix} = \begin{bmatrix} \text{---} a_1^* \text{---} \\ \text{---} a_2^* \text{---} \\ \vdots \\ \text{---} a_m^* \text{---} \end{bmatrix} \begin{Bmatrix} x \end{Bmatrix}$$

so that

$$y_i = \sum_{j=1}^n \underbrace{\boxed{a_i(j)}}_{A_{ij}} x_j = \sum_{j=1}^n A_{ij} x_j$$

which is again the classical matrix vector multiplication formula.

- THESE SETS OF INNER PRODUCTS CAN BE VIEWED AS A WAY TO DEFINE A PARTICULAR MAP, SO THIS DEFINITION DOES NOT CONTRADICT THE MOST GENERAL ONE THAT VIEWS A MATRIX-VECTOR MULTIPLICATION AS A LINEAR FUNCTION.