

Linear Algebra II

① Review LA I & Motivation

What did we do in LA I ?

- Solve linear systems
- Matrices & vectors

$$Ax = b$$

$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n, \quad A \in \mathbb{R}^{m \times n}$$

$$b \in \mathbb{R}^m$$

for these we had some operations

$$\begin{pmatrix} 1 \\ 2 \end{pmatrix} + \begin{pmatrix} 3 \\ 4 \end{pmatrix} = \begin{pmatrix} 4 \\ 6 \end{pmatrix} \quad \bullet \text{ Addition } x + y \in \mathbb{R}^n \quad x, y \in \mathbb{R}^n$$

$$3 \cdot \begin{pmatrix} 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 3 \\ 6 \end{pmatrix} \quad \bullet \text{ Scalar multiplication } \lambda x \in \mathbb{R}^n \quad \lambda \in \mathbb{R}$$

- Linear maps $F: \mathbb{R}^n \rightarrow \mathbb{R}^m$

$$i) F(x+y) = F(x) + F(y)$$

$$ii) F(\lambda x) = \lambda F(x)$$

- Subspaces $U \subset \mathbb{R}^n$ $0 = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix} \in \mathbb{R}^n$

$$i) 0 \in U$$

$$ii) x, y \in U \Rightarrow x + y \in U$$

$$iii) x \in U \Rightarrow \lambda x \in U$$

- Image & Kernel of linear maps

$$F: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$\underbrace{\quad}_{\text{ker } F} \quad \underbrace{\quad}_{\text{im } F}$$

- Linear independency & Basis

In "real life" there are a lot of other examples of objects with "Addition" and "Scalar multiplication" for which these concepts are interesting.

Example 1

Let $\mathcal{F}(\mathbb{R}, \mathbb{R})$ be the set of all functions $\mathbb{R} \rightarrow \mathbb{R}$.

$e^x \sin(x) x^2+3$

This is a really big set!

For $f, g \in \mathcal{F}(\mathbb{R}, \mathbb{R})$ we can also define $f+g$ and λf ($\lambda \in \mathbb{R}$),

by $(f+g)(x) = f(x) + g(x)$ and $(\lambda f)(x) = \lambda \cdot f(x)$.

Now consider the subset $U = \{f \in \mathcal{F}(\mathbb{R}, \mathbb{R}) \mid f'' = f\}$.
we have:

i) The function $n(x) = 0$ is an element of U .

ii) If $f, g \in U$, i.e. $f'' = f$ and $g'' = g$ then

$$(f+g)'' = f'' + g'' = f + g \Rightarrow f+g \in U.$$

iii) $\lambda \in \mathbb{R}, f \in U \Rightarrow \lambda f \in U$.

Therefore: U can be seen as a subspace of $\mathcal{F}(\mathbb{R}, \mathbb{R})$.

(The "zero vector" $\begin{pmatrix} 0 \\ 0 \end{pmatrix}$ gets replaced by the "zero function" $n(x) = 0$ $\forall x \in \mathbb{R}$).

One can show that $f_1(x) = e^x, f_2(x) = e^{-x}$ and their linear combinations $\lambda_1 f_1 + \lambda_2 f_2$ are the only elements in U . ($\lambda_1, \lambda_2 \in \mathbb{R}$)
($U = \text{span}\{f_1, f_2\}$)

Moreover: f_1 and f_2 are linearly independent, i.e.

$$\lambda_1 f_1(x) + \lambda_2 f_2(x) = 0 \quad \forall x \in \mathbb{R} \Rightarrow \lambda_1 = \lambda_2 = 0.$$

$$(\lambda_1 f_1 + \lambda_2 f_2 = n)$$

$\Rightarrow (f_1, f_2)$ is a basis of $U \Rightarrow \dim U = 2$.

① Vector spaces

Spaces for which "Addition" and "Scalar multiplication" are defined in such a way that they satisfy the same computation rules as vectors are called vector spaces.

Definition 1.1 A (real) vector space is a set V together with two functions

$$+ : V \times V \longrightarrow V$$

$$(u, v) \longmapsto u+v$$

Addition

$$\cdot : \mathbb{R} \times V \longrightarrow V$$

$$(\lambda, v) \longmapsto \lambda \cdot v = \lambda v$$

Scalar multiplication

satisfying:

Properties for the addition:

$$(A.1) \quad \forall u, v, w \in V : (u+v)+w = u+(v+w) \quad (\text{Associativity})$$

$$(A.2) \quad \forall u, v \in V : u+v = v+u \quad (\text{Commutativity})$$

$$(A.3) \quad \text{There exist an element } n \in V \text{ (Identity/neutral element), such that } \forall u \in V : n+u = u.$$

$$(A.4) \quad \text{Every element } u \in V \text{ has an additive inverse, i.e. there exist a } v \in V \text{ with } u+v = n.$$

Compatibility of addition and scalar multiplication:

$$(C.1) \quad \forall u, v \in V, \lambda \in \mathbb{R} : \lambda(u+v) = \lambda u + \lambda v$$

$$(C.2) \quad \forall u \in V, \lambda, \mu \in \mathbb{R} : (\lambda+\mu)u = \lambda u + \mu u$$

$$(C.3) \quad \forall u \in V, \lambda, \mu \in \mathbb{R} : \lambda(\mu u) = (\lambda\mu)u$$

$$(C.4) \quad \forall u \in V : 1u = u.$$

Example 2 i) $V = \mathbb{R}^n$ with usual operations

$$0 = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}, \quad u = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}, \quad -u = \begin{pmatrix} -u_1 \\ \vdots \\ -u_n \end{pmatrix}$$

ii) $\mathcal{F}(\mathbb{R}, \mathbb{R}) = \{f: \mathbb{R} \rightarrow \mathbb{R}\}$

$$(f+g)(x) = f(x) + g(x), \quad (\lambda f)(x) = \lambda \cdot f(x) \quad \forall x \in \mathbb{R}$$

$$n(\ast) = 0 \quad \forall x, \quad (-f)(x) = -f(x).$$

$$\mathcal{C}^0(\mathbb{R}, \mathbb{R}) = \{f \in \mathcal{F}(\mathbb{R}, \mathbb{R}) \mid f \text{ is continuous}\}.$$

iii) Infinite sequences $(a_n)_{n \geq 1} = (a_1, a_2, \dots)$

\mathcal{J} = set of all infinite sequences.

iv) Matrices: $V = \mathbb{R}^{m \times n}$ with usual addition & scalar mult.

$$0 = \begin{pmatrix} 0 & \dots & 0 \\ \vdots & & \vdots \\ 0 & \dots & 0 \end{pmatrix}$$

In the following V always denotes a vector space

Definition 1.3 A subset $U \subset V$ is a subspace (of V) if

i) $0 \in U$

ii) $\forall u, v \in U : u + v \in U$

iii) $\forall u \in U, \lambda \in \mathbb{R} : \lambda u \in U.$

Example 3 i) $\mathcal{P} = \{f: \mathbb{R} \rightarrow \mathbb{R} \mid f \text{ is a polynomial fct.}\}$
 e.g. $f(x) = 2x^3 - x^2 + 1$

$$\mathcal{P} \subset \mathcal{C}^0(\mathbb{R}, \mathbb{R}) \subset \mathcal{F}(\mathbb{R}, \mathbb{R})$$

subspace

\cup
 U (from Example 1)

ii) Subspaces of \mathbb{R}^n

iii) $\{(a_n)_{n \geq 1} \in \mathcal{J} \mid \lim_{n \rightarrow \infty} a_n \text{ exists}\} \subset \mathcal{J}$

e.g. $a_n = \frac{1}{n}$

iv) $\{\text{invertible matrices}\} \subset \mathbb{R}^{m \times n}$ are Not a subspace

Proposition 1.4 If $U \subset V$ is a subspace then U is also a vector space with the operations inherited from V .

Proof: Exercise (follows directly from the definition).

Definition 1.5 V vector space, $v_1, \dots, v_n \in V$

i) Span of v_1, \dots, v_n : $\text{span}\{v_1, \dots, v_n\} = \left\{ \underbrace{\sum_{i=1}^n \lambda_i v_i}_{\text{linear combination of } v_1, \dots, v_n} \mid \lambda_1, \dots, \lambda_n \in \mathbb{R} \right\}$

ii) v_1, \dots, v_n span (or generate) V if $\text{span}\{v_1, \dots, v_n\} = V$.
 V is finitely generated if there exist v_1, \dots, v_n s.t. $V = \text{span}\{v_1, \dots, v_n\}$

iii) v_1, \dots, v_n are linearly independent if $\lambda_1 v_1 + \dots + \lambda_n v_n = 0 \Rightarrow \lambda_1 = \dots = \lambda_n = 0$

iv) $B = (v_1, \dots, v_n)$ is a basis of V if v_1, \dots, v_n are lin. ind. and $\text{span}\{v_1, \dots, v_n\} = V$.

Convention: $\{0\} = \text{span}\{\emptyset\}$ is also ^{empty set} fin. gen.

Example 4

i) $\mathbb{R}^n = \text{span}\{e_1, \dots, e_n\}$ is fin. gen.

ii) $\mathcal{P} = \left\{ f: \mathbb{R} \rightarrow \mathbb{R} \mid f(x) = \sum_{i=1}^m a_i x^i, a_1, \dots, a_m \in \mathbb{R} \right\}$

is not fin. gen.:

If $f_1, \dots, f_m \in \mathcal{P}$ would generate \mathcal{P} , then $g(x) = x^{d+1}$ would not be in $\text{span}\{f_1, \dots, f_m\}$ if $d = \max\{\deg(f_i) \mid i=1, \dots, m\}$.

iii) Set $\mathcal{P}_n = \{f \in \mathcal{P} \mid \deg(f) \leq n\}$. \mathcal{P}_n is finitely generated with basis $B = (f_0, \dots, f_n)$, where $f_j(x) = x^j$.

The following Propositions, Lemmas & Theorems were already proven in LA I for the case $V = \mathbb{R}^n$. Almost all of their proofs are exactly the same.

Proposition 1.6 $v_1, \dots, v_n \in V$. The following statements are equivalent.

- i) v_1, \dots, v_n are lin. dependent (i.e. $\exists \lambda = \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_n \end{pmatrix} \neq 0$ with $\sum_{i=1}^n \lambda_i v_i = 0$)
- ii) $\exists j : v_j \in \text{span}\{v_1, \dots, \cancel{v_j}, \dots, v_n\}$
- iii) $\exists j : \text{span}\{v_1, \dots, v_n\} = \text{span}\{v_1, \dots, \cancel{v_j}, \dots, v_n\}$

Proof: Same as for $V = \mathbb{R}^n$

Lemma 1.7 If $v_1, \dots, v_l \in V$ are lin. indep. and $V = \text{span}\{w_1, \dots, w_m\}$ then $l \leq m$

Proof: We will show that if $l > m$, then v_1, \dots, v_l are lin. dependent.

For $j = 1, \dots, l$ write $v_j = \sum_{i=1}^m \alpha_{ij} w_i$ ($\alpha_{ij} \in \mathbb{R}$)

We have
$$\sum_{j=1}^l \lambda_j v_j = \sum_{j=1}^l \lambda_j \left(\sum_{i=1}^m \alpha_{ij} w_i \right) = \sum_{i=1}^m \left(\underbrace{\sum_{j=1}^l \lambda_j \alpha_{ij}}_{=: \beta_i} \right) w_i$$

Set $A = (\alpha_{ij}) = \begin{pmatrix} \alpha_{11} & \dots & \alpha_{1l} \\ \vdots & & \vdots \\ \alpha_{m1} & \dots & \alpha_{ml} \end{pmatrix} \in \mathbb{R}^{m \times l}$, $\lambda = \begin{pmatrix} \lambda_1 \\ \vdots \\ \lambda_l \end{pmatrix}$

This gives $A\lambda = \begin{pmatrix} \beta_1 \\ \vdots \\ \beta_m \end{pmatrix}$.

If $l > m$ the equation $A\lambda = \begin{pmatrix} 0 \\ \vdots \\ 0 \end{pmatrix}$ has infinitely many solutions.

In particular $\exists \lambda \neq 0$ with $A\lambda = 0 \Rightarrow \sum_{j=1}^l \lambda_j v_j = 0 \Rightarrow v_1, \dots, v_l$ lin. dependent \square

Theorem 1.8 Let V be finitely generated.

- i) V has a (finite) basis
- ii) All bases of V have the same number of elements
- iii) If $v_1, \dots, v_k \in V$ are linearly independent then there exist $v_{k+1}, \dots, v_n \in V$ such that (v_1, \dots, v_n) is a basis of V .
- iv) If $V = \text{span}\{w_1, \dots, w_m\}$, then there exist a subset $\{u_1, \dots, u_k\} \subset \{w_1, \dots, w_m\}$, such that (u_1, \dots, u_k) is a basis of V .

Proof: Same as in LA I. (ii) follows from Lemma 1.7
(i) follows from iv)

Definition 1.9 V fin. gen. with basis (v_1, \dots, v_n) . Then $\dim(V) = n$ is the dimension of V .

Corollary 1.10 Let $\dim V = n$, $v_1, \dots, v_n \in V$. Then the following statements are equivalent:

- i) v_1, \dots, v_n are lin. indep.
- ii) $V = \text{span}\{v_1, \dots, v_n\}$
- iii) (v_1, \dots, v_n) is a basis of V .

Proposition 1.11 Let V be fin. gen. and $U \subset V$ be a subspace. Then U is also finitely generated.

Proof: Can be done by using Prop. 1.6 + Lemma 1.7.

