

Vv285 Recitation Class 1

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Outline

- 1 About Me and My RC
- 2 Review of Vector Spaces
- 3 Linear Equations & Basis
- 4 Inner Product
- 5 Reference

Logistics and Style

- Mind-Map track through Horst's slides
 - ① More intuitive explanation.
 - ② No repetition. Ideal: my RC = (Horst' slides)[⊥].
 - ③ Marked importance.
- Exercises with emphasis
 - ① 80% of them is close to exam's difficulty
 - ② 20% of them is a bit harder, through which I hope to strengthen your understanding for some important concepts
- Interactions and timely-feedback
 - ① Hope to see your faces online :)
 - ② Welcome to any questions during or after my RC
 - ③ Don't be shy posting your solution on the chat board or just speaking up.

Vector Space

Definition: A triple $(V, +, \cdot)$ is called a **real** vector space if

1. V is any set
2. $+: V \times V \rightarrow V$ is a map (addition) with the following properties:
 - Communicativity:
 $u + v = v + u$ for all $u, v \in V$
 - Associativity:
 $(u + v) + w = u + (v + w)$ for all $u, v, w \in V$
 - Additive Identity:
 there exist an element $0 \in V$ such that $v + 0 = v$ for all $v \in V$
 - Additive Inverse:
 for every $v \in V$, there exists $w \in V$ such that $v + w = 0$

Vector Space

3. $\cdot : \mathbb{R} \times V \rightarrow V$ is a map (scalar multiplication) with the following properties:
- Multiplication Identity:
 $1 \cdot u = u$ for all $u \in V$
 - Associativity:
 $(ab)v = a(bv)$ for all $u, v, w \in V$ and all $a, b \in \mathbb{R}$
 - Distributive Property:
 $a(u + v) = au + av$ and $(a + b)v = av + bv$ for all $a, b \in \mathbb{R}$ and all $u, v \in V$

Subspace

Definition: Let $(V, +, \cdot)$ be a real or complex vector space. If $U \subseteq V$ and $(U, +, \cdot)$ is also a vector space, then we say that $(U, +, \cdot)$ is a subspace of $(V, +, \cdot)$

Subspace

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There is no need to prove the above properties of addition and scalar multiplication of $(U, +, \cdot)$, actually, we only need to verify that

- $0 \in U$
- $u_1 + u_2 \in U$ for all $u_1, u_2 \in U$
- $\lambda u \in U$ for all $u \in U$ and $\lambda \in \mathbb{R}$ (or $\lambda \in \mathbb{C}$ for complex vector space)

In a word, we need to verify if this subset is closed under addition and multiplication, with confirmation of unit element.

you can review the proof on Vv186 *Slide 339*

Exercise - Is this a vector space

Let ∞ and $-\infty$ denote two distinct objects, neither of which is in \mathbf{R} . Define an addition and scalar multiplication on $\mathbf{R} \cup \{\infty\} \cup \{-\infty\}$ as you could guess from the notation. Specifically, the sum and product of two real numbers is as usual, and for $t \in \mathbf{R}$ define

$$t\infty = \begin{cases} -\infty & \text{if } t < 0, \\ 0 & \text{if } t = 0, \\ \infty & \text{if } t > 0, \end{cases} \quad t(-\infty) = \begin{cases} \infty & \text{if } t < 0, \\ 0 & \text{if } t = 0, \\ -\infty & \text{if } t > 0, \end{cases}$$

$$t + \infty = \infty + t = \infty, \quad t + (-\infty) = (-\infty) + t = -\infty,$$

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Is $\mathbf{R} \cup \{\infty\} \cup \{-\infty\}$ a vector space over \mathbf{R} ? Explain.

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No! Uniqueness of unit element, i.e, **0**

Real and Complex Vector Spaces

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- The set of n-tuples (n dimensional vectors)
 $\mathbb{R}^n = \{x = (x_1, \dots, x_n) : x_1, \dots, x_n \in \mathbb{R}\}$ is a real vector space if the addition is defined by

$$\begin{aligned} x + y &= (x_1, \dots, x_n) + (y_1, \dots, y_n) \\ &:= (x_1 + y_1, \dots, x_n + y_n), \quad x, y \in \mathbb{R} \end{aligned}$$

and scalar multiplication defined by

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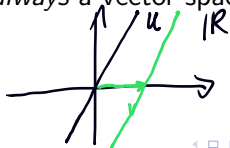
- The above vector space is complex if the multiplication factor λ is defined over \mathbb{C} .

Exercises

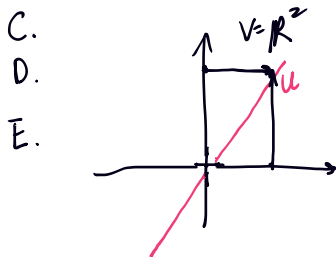
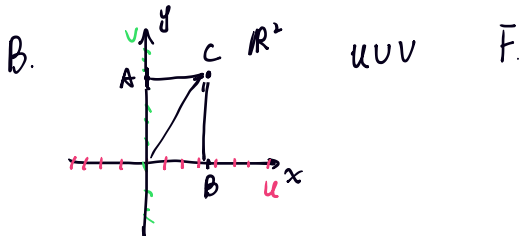
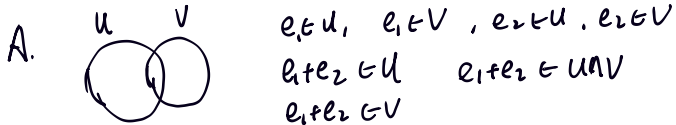
V, U are real (complex) vector spaces, then which of the followings are true

- A $V \cap U$ is a vector space
- B $V \cup U$ is a vector space
- C $V + U := \{u + v : \forall u \in U, \forall v \in V\}$ is a vector space
- D Suppose U is a subspace of V , then $V \setminus U$ is a vector space
- E Suppose U is a subspace of V , then $(V \setminus U) \cup \{0\}$ is a vector space
- F Suppose U is a subspace of V and $v \in V$, then $v + U := \{v + u : u \in U\}$ is *always* a vector space

affine set



Quotient Space



Exercises

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Answer: A and C

Normed Vector Space

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Norm is a map (function) defined over a vector space V , $\|\cdot\| : V \rightarrow \mathbb{R}$, such that for all $u, v \in V$ and all $\lambda \in \mathbb{R}$ (\mathbb{C} if V is a complex vector space):

1. $\|v\| \geq 0$ and $\|v\| = 0$ if and only if $v = 0$
2. $\|\lambda \cdot v\| = |\lambda| \cdot \|v\|$
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If we can define a norm over a vector space V , then we say V is a **normed vector space**.

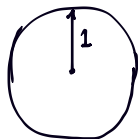
Normed Vector Spaces - Examples

➤ p-Norm: $\|x\|_p = \left(\sum_{j=1}^n |x_j|^p\right)^{1/p}$ for any $p \in \mathbb{N} \setminus \{0\}$ in \mathbb{R}^n

when $p = 2$ we obtain the usual Euclidean norm

when $p = \infty$ obtain the max norm $\|x\|_\infty = \max_{1 \leq k \leq n} |x_k|$

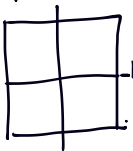
$p=2$



$p=1$



$p=\infty$



Normed Vector Spaces - Examples

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➤ $C([a, b])$, $[a, b] \subseteq \mathbb{R}$, with $\|f\|_\infty = \sup_{x \in [a, b]} |f(x)|$.

➤ Later we'll talk about inner product space and induced norm.

Review for this week - Linear equations & Basis

1. Systems of linear equations - know the proof of fundamental lemma for homogeneous equations
2. Finite dimensional vector space
 - Linear combination and span
 - **Important:** definition of basis (unique presentation), property (1. span the whole space 2. linear independence)
 - Dimension of vector space
 - Basis extension theorem
 - Sums of vector spaces. **Important:** direct sum

Something Horst asks us to do

3.28. **Theorem.** Let V be a vector space and $U, W \subset V$ be finite-dimensional subspaces of V . Then

$$\dim(U + W) + \dim(U \cap W) = \dim U + \dim W.$$

The proof will be discussed in recitation class.



①. $U \cap W = \{0\}$

② non-trivial

$$U \cap W : B_1 = \{e_1, \dots, e_n\}$$

$$U = \{e_1, \dots, e_n\} \cup B_2 = \{v_1, \dots, v_m\}$$

$$W = \{e_1, \dots, e_n\} \cup B_3 = \{b_1, \dots, b_j\}$$

$$\sum_i \lambda_i e_i + \sum_k \mu_k v_k + \sum_p \alpha_p b_p = 0$$

$$\underbrace{\sum_i \lambda_i e_i}_{\in U} + \underbrace{\sum_k \mu_k v_k}_{\in U} = \underbrace{\left(\sum_p \alpha_p b_p \right)}_{\in W} \quad \alpha_p = 0 \quad \downarrow$$

$U \cap W$

$$B_0 = \{v_1, \dots, v_m, b_1, \dots, b_j\} \cup \{e_1, \dots, e_n\}$$

Review for this week - Inner product

Inner Product Space - Basic properties

- Positivity
- Definiteness
- Additivity in the **second** slot (note the difference with other references!)
- Homogeneity in the **second** slot
- Conjugate symmetry

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How to prove the additivity in the first slot ?

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How to prove the additivity in the first slot ?

Is inner product linear in the first slot ?

Review for this week - Inner product

➤ Cauchy-Schwarz Inequality: $|\langle u, v \rangle| \leq \|u\| \cdot \|v\|$



Review for this week - Inner product

- Cauchy-Schwarz Inequality: $|\langle u, v \rangle| \leq \|u\| \cdot \|v\|$
More intuitive proof – using orthogonal projection
- Induced norm. Question: *Is every norm an induced norm?*
- Angle between vectors. $\{e_1, \dots, e_n\}$ $\textcircled{1} V = \sum_k \langle e_k, v \rangle e_k$ $\textcircled{2} V = \sum_k \langle v, e_k \rangle e_k$ $\textcircled{1}$
- Important: **Orthogonality**, orthogonal complement, Pythagoras' Theorem (the generality of linear algebra) $\textcircled{2} V = \sum_k \langle v, e_k \rangle e_k$ $\textcircled{1}$
- Orthonormal bases (orthogonal + normal (unit norm)), Parseval's Theorem
- Projection theorem: decompose a vector in to U and U^\perp , which is *unique*
- Bessel's inequality

$$\langle e_k, v \rangle = \lambda_k e_k$$

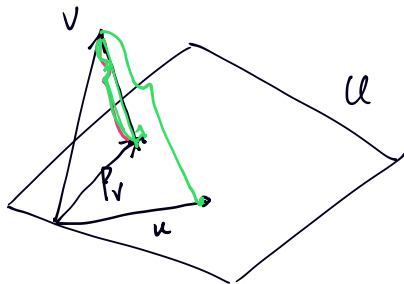
Review for this week - Inner product

- Simpler proof for best approximation (slides 104)

Suppose P_v is the projection from $v \in V$ to U , i.e., $P_v = \sum_{i=1}^r \lambda_i e_i$, and

$u \in U$ is any vector.

We want to show $\|v - P_v\| \leq \|v - u\|$



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We want to show $\|v - P_v\| \leq \|v - u\|$

Proof:

$$\begin{aligned} \|v - u\|^2 &= \left\| \underbrace{v - P_v}_{\in U^\perp} + \underbrace{P_v - u}_{\in U} \right\|^2 = \|v - P_v\|^2 + \|P_v - u\|^2 \\ &\geq \|v - P_v\|^2 \end{aligned}$$

- Important: **Gram-Schmidt Orthonormalization** (remember to reduce the norm to unit!)

Exercises - Orthogonal Complement

Suppose U is a subspace for a finite dimension vector space V , prove that

$$U = (U^\perp)^\perp$$

$$\begin{cases} U \oplus (U^\perp) = V \\ (U^\perp)^\perp = V \end{cases}$$

$$U = (U^\perp)^\perp$$

$$\textcircled{1} U \subseteq (U^\perp)^\perp$$

$$\boxed{u \in U}, \boxed{\langle u, v \rangle = 0}, v \in U^\perp$$

$$\textcircled{2} U \supseteq (U^\perp)^\perp$$

$$v \in (U^\perp)^\perp$$

$$v = u + w$$

$$u \in U, w \in U^\perp$$

$$v - u = w \in U^\perp$$

$$v - u \in (U^\perp)^\perp$$

$$v - u \in (U^\perp) \cap (U^\perp)^\perp = \{0\}$$

$$v = u$$

$$\Rightarrow (U^\perp)^\perp \subseteq U$$

Exercise* - deterministic property of inner product

Let V be a real vector space and $\tau : V \times V \rightarrow \mathbb{R}$. Show that if

(a) $\tau(v, v) > 0$ for **some** $v \in V$;

(b) $\tau(v, v) \neq 0$ for any $v \neq 0$;

(c) $\tau(v, \lambda u)$ $= \lambda \tau(v, u)$ for $v, u \in V$ and $\lambda \in \mathbb{R}$;

(d) $\tau(v, u)$ $= \tau(u, v)$;

(e) $\tau(v, u + w) = \tau(v, u) + \tau(v, w)$

then τ is an inner product on V .

$$|x - y| < \delta \Rightarrow |f(x) - f(y)| < \epsilon$$

$$|x - \lambda x| < \delta$$

Exercise* - deterministic property of inner product

Proof:¹

$$\tau(v, v) > 0$$

Suppose that there is some $u \in V$ with $\tau(u, u) < 0$. Let $g: [0, 1] \rightarrow \mathbb{R}$ be defined by

$$g(t) = \tau(tv + (1 - t)u, tv + (1 - t)u).$$

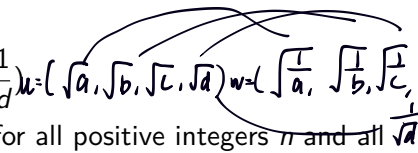
g is continuous because of (c) and (d). By the Mean Value Theorem for continuous functions, $g(t_0) = 0$ for some $t_0 \in (0, 1)$. Since $v \neq t^{-1}(t - 1)u$ (why?), this shows that $\tau(h, h) = 0$ for some $h \neq 0$, contradicting (b).

Therefore, $\tau(h, h) > 0$ for all $h \in V \setminus \{0\}$. Then (c) implies that $\tau(0, 0) = 0$. It follows that τ is an inner product on V .

¹This proof is from Leyang Zhang

Exercise - Cauchy-Schwarz Inequality

Prove that

1. $16 \leq (a + b + c + d)\left(\frac{1}{a} + \frac{1}{b} + \frac{1}{c} + \frac{1}{d}\right)$ 
2. $(x_1 + \cdots + x_n)^2 \leq n(x_1^2 + \cdots + x_n^2)$ for all positive integers n and all real numbers x_1, \dots, x_n

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$$\| \langle u, v \rangle \|^2 \leq \|u\|^2 \|v\|^2$$

$$(x_1 + \cdots + x_n)^2 \leq n(x_1^2 + \cdots + x_n^2)$$

Hint

- ① for 2, try $u = (1, \dots, 1)$ and $v = (x_1, \dots, x_n)$

Exercise - Gram-Schmidt Orthonormalization

On $\mathcal{P}_2(\mathbf{R})$, consider the inner product given by

$$\langle p, q \rangle = \int_0^1 p(x)q(x) dx.$$

Apply the Gram-Schmidt Procedure to the basis $1, x, x^2$ to produce an orthonormal basis of $\mathcal{P}_2(\mathbf{R})$.

Exercise - Gram-Schmidt Orthonormalization

Solution:

Memorize the algorithm, given $\{v_1, v_2, v_3\}$ an independent set:

$$e_1 = \frac{v_1}{\|v_1\|} = 1$$

$$e_2 = \frac{v_2 - \langle v_2, e_1 \rangle e_1}{\|v_2 - \langle v_2, e_1 \rangle e_1\|} = 2\sqrt{3}\left(x - \frac{1}{2}\right)$$

$$e_3 = \frac{v_3 - \langle v_3, e_2 \rangle e_2 - \langle v_3, e_1 \rangle e_1}{\|v_3 - \langle v_3, e_2 \rangle e_2 - \langle v_3, e_1 \rangle e_1\|} = 6\sqrt{5}\left(x^2 - x + \frac{1}{6}\right)$$

References I

- VV186 slides from Horst Hohberger
- VV285 slides from Horst Hohberger
- Answers from Leyang Zhang
- Linear Algebra Done Right from Axler