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Lecture No. Inner product

1. Diagonalization of linear maps.

We recall that λ is called an eigenvalue of a linear operator $T: V \rightarrow V$ if

$$Tv = \lambda v \quad (9.1)$$

for some non-zero vector $v \in V$.

The vector v in (9.1) is called an eigenvector corresponding to λ .

Th 9.4 A linear map $T: V \rightarrow V$ is diagonalizable iff $\exists v_1, \dots, v_n$ -basis in V such that v_i are eigenvectors corresponding to λ_i .

• In this case

$$M_T = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix}$$

in this basis

Th 9.1 Let $\dim V = n$ and T have n distinct eigenvalues, then T is diagonalizable.

(2) Proof. Let $\lambda_1, \dots, \lambda_n$ be distinct eigenvalues of T . Then there exists v_1, \dots, v_n - eigen vectors corresponding to $\lambda_1, \dots, \lambda_n$. By Th 8.3, they are linearly independent. Thus, v_1, \dots, v_n form a basis in V .

By Th 8.4. T is diagonalizable and

$$M_T = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{pmatrix}$$

in the basis v_1, \dots, v_n .

Ex. 91 $A = \begin{pmatrix} 1 & 1 \\ 4 & 1 \end{pmatrix} \in \mathbb{R}^{2 \times 2}$

$$\det(A - \lambda I) = \begin{vmatrix} 1-\lambda & 1 \\ 4 & 1-\lambda \end{vmatrix} = (1-\lambda)^2 - 4 =$$

$$= (1-\lambda-2)(1-\lambda+2) = (-1-\lambda)(3-\lambda) = 0$$

$$\lambda_1 = -1, \lambda_2 = 3$$

Next we find eigen vectors:

$$\lambda_1: A v = \lambda_1 v \Leftrightarrow (A - \lambda_1 I) v$$

$$\begin{pmatrix} 1+1 & 1 \\ 4 & 1+1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 0$$

$$\begin{cases} 2x + y = 0 \\ 4x + 2y = 0 \end{cases} \Rightarrow \begin{cases} x = 1 \\ y = -2 \end{cases} \quad v_1 = (1, -2)$$

$$\lambda_2: \begin{pmatrix} -2 & 1 \\ 4 & -2 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = 0 \Rightarrow \begin{cases} -2x + y = 0 \\ 4x - 2y = 0 \end{cases} \Rightarrow \begin{cases} x = 1 \\ y = 2 \end{cases}, \quad v_2 = (1, 2)$$

(3) Thus, A is diagonal in basis $v_1 = (1, -2)$
 $v_2 = (1, 2)$.

Th 9.2 Let $\dim V = n$, $T: V \rightarrow V$ be a linear map, $\lambda_1, \dots, \lambda_m$ be distinct eigenvalues of T and

$$V_{\lambda_i} = \text{Ker}(T - \lambda_i I) = \{v : Tv = \lambda_i v\}$$

be corresponding eigenspaces.

The following conditions are equivalent:

- (i) T is diagonalizable
- (ii) The characteristic polynomial of T is

$$\det(M_T - \lambda I) = (\lambda_1 - \lambda)^{n_1} \dots (\lambda_m - \lambda)^{n_m}$$

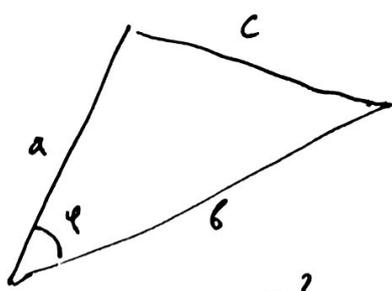
and $\dim V_{\lambda_i} = n_i, i=1, \dots, m$

(iii) $\dim V_{\lambda_1} + \dots + \dim V_{\lambda_m} = n$.

2. Scalar product in \mathbb{R}^3 (in \mathbb{R}^2)

In this section we will work with the space \mathbb{R}^2 . All results can be easily transferred to the case of \mathbb{R}^3 .

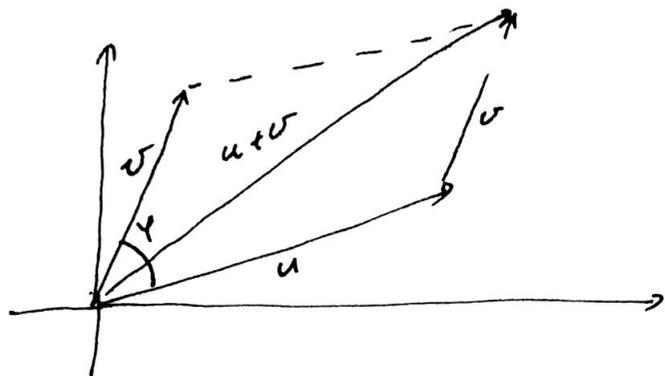
Let us consider a triangle :



Law of cosines: $c^2 = a^2 + b^2 - 2ab \cos \varphi$

(4)

Now we consider two vectors



Then $|u+v|^2 = |v|^2 + |u|^2 + 2|u||v| \cdot \overbrace{\cos \varphi}^{\text{scalar product}}$,
 where φ is the angle between v and u .

● Def 9.1 The number $(u, v) = |u||v| \cos \varphi$
 is called the scalar product (dot product)
 of vectors u and v .

Th 9.3 Let $u, v, w \in \mathbb{R}^3$ (\mathbb{R}^2), $a \in \mathbb{R}$. Then

1) $(u+v, w) = (u, w) + (v, w)$, $(au, v) = a(u, v)$

2) $(u, u) = |u|^2 \geq 0$, $|u| = \sqrt{(u, u)}$

3) $(u, u) = 0$ iff $u = 0$

4) $(u, v) = (v, u)$

5) $(u, v) = 0$ iff u, v are orthogonal

6) $(u, v) = a_1 b_1 + a_2 b_2 + a_3 b_3$, if $u = (a_1, a_2, a_3)$

7) $\cos \varphi = \frac{(u, v)}{|u||v|}$, where φ is the
 angle between u, v .
 $v = (b_1, b_2, b_3)$

⑥ Proof we only check 6) since other properties are trivial.

We compute

$$\begin{aligned} |u+v|^2 &= |(a_1+b_1, a_2+b_2, a_3+b_3)|^2 = \\ &= (a_1+b_1)^2 + (a_2+b_2)^2 + (a_3+b_3)^2 = \\ &= \underbrace{a_1^2 + a_2^2 + a_3^2}_{|u|^2} + \underbrace{b_1^2 + b_2^2 + b_3^2}_{|v|^2} + 2 \underbrace{(a_1 b_1 + a_2 b_2 + a_3 b_3)}_{(u,v)}. \quad \square \end{aligned}$$

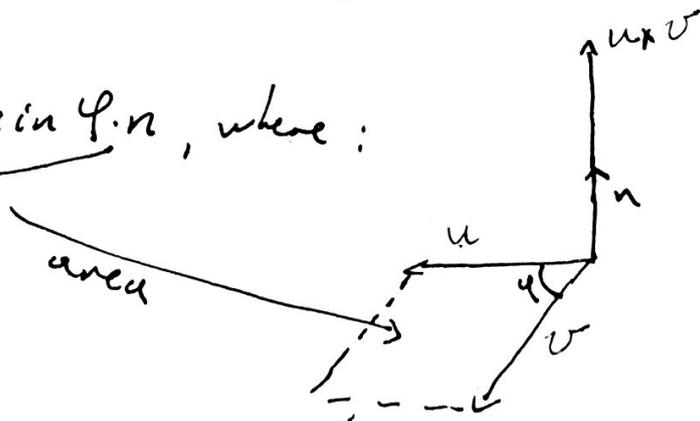
Def 9.2 Vectors v_1, v_2, v_3 in \mathbb{R}^3 are called an orthonormal basis in \mathbb{R}^3 if they are orthogonal and have length 1, that is

$$(v_i, v_j) = \delta_{ij} = \begin{cases} 1, & i=j \\ 0, & i \neq j \end{cases}$$

3 Vector product in \mathbb{R}^3

Def 9.3 The vector product (cross product) $u \times v$ is defined as a vector w that is orthogonal to u and v , with a direction given by the right hand rule and the length of the area of the parallelogram that the vectors span.

$$u \times v = \underbrace{|u||v| \sin \varphi}_{\text{area}} \cdot n, \text{ where:}$$



⑥

We note that

$$u \times v = -v \times u.$$

Th 9.4 Let $u, v, w \in \mathbb{R}^3$ and $a \in \mathbb{R}$. Then

1) $u \times v = 0$ iff u, v are collinear (parallel)

2) $(u+v) \times w = u \times w + v \times w$

3) $(au) \times v = a(u \times v)$

4) If i, j, k be an oriented orthonormal basis, i.e. they form an orthonormal basis and $i \times j = k$, then

$$u \times v = \begin{vmatrix} i & j & k \\ a_1 & a_2 & a_3 \\ b_1 & b_2 & b_3 \end{vmatrix} = \begin{vmatrix} a_2 & a_3 \\ b_2 & b_3 \end{vmatrix} i - \begin{vmatrix} a_1 & a_3 \\ b_1 & b_3 \end{vmatrix} j +$$

for $u = a_1 i + a_2 j + a_3 k$
 $v = b_1 i + b_2 j + b_3 k$

$$+ \begin{vmatrix} a_1 & a_2 \\ b_1 & b_2 \end{vmatrix} k$$

5) $(u, v \times w) = (v, w \times u) = (w, u \times v)$.

6) $u \times (v \times w) = v(u, w) - w(u, v)$

4. Inner product

Let V be a vector space over \mathbb{F} .

Def. 9.4 An inner product on V is a map

$$\langle \cdot, \cdot \rangle: V \times V \rightarrow \mathbb{F}$$

satisfying the following properties

1) Linearity in first slot: $\langle u+v, w \rangle = \langle u, w \rangle + \langle v, w \rangle$

$$\langle au, v \rangle = a \langle u, v \rangle,$$

for all $u, v, w \in V$ and $a \in \mathbb{F}$.

(7)

2) Positivity: $\langle v, v \rangle \geq 0$ for all $v \in V$;

3) Positive definiteness: $\langle v, v \rangle = 0$ iff $v = 0$;

4) Conjugate symmetry: $\langle u, v \rangle = \overline{\langle v, u \rangle}$ for all $u, v \in V$

Remark. $\langle u, \alpha v \rangle = \alpha \langle u, v \rangle$ (if $F = \mathbb{R}$, then $\langle u, v \rangle = \langle v, u \rangle$)

Def 9.5 An inner product space is a vector space over F together with an inner product $\langle \cdot, \cdot \rangle$.

Ex 9.2 Let $V = F^n$, $u = (a_1, \dots, a_n)$, $v = (b_1, \dots, b_n)$

Then $\langle u, v \rangle = \sum_{i=1}^n a_i b_i$

if $F = \mathbb{R}$, then $\langle u, v \rangle$ is the usual scalar product $\langle u, v \rangle = a_1 b_1 + \dots + a_n b_n$

Ex 9.3 $V = \mathbb{F}[z]$ or $V = C([0, 1])$

$\langle f, g \rangle = \int_0^1 f(z) \overline{g(z)} dz$

Def 9.6 A map

$\|\cdot\| : V \rightarrow [0, +\infty)$

is a norm on V if

1) Positive homogeneity: $\| \alpha v \| = |\alpha| \|v\| \quad \forall \alpha \in F, v \in V$;

2) Positive definiteness: $\|v\| = 0$ iff $v = 0$;

3) Triangle inequality $\|v + w\| \leq \|v\| + \|w\| \quad \forall v, w \in V$.

⑧

If V is an inner product space, we can define

$$\|u\| := \sqrt{\langle u, u \rangle}. \quad (9.2)$$

It turns out that $\|\cdot\|$ is a norm on V .

Indeed, conditions 1) 2) of Def 9.6 trivially follows from definition of the inner product. We need to check only triangular inequality. First, we prove the following theorem:

Th 9.5 (Cauchy - Schwarz inequality). Let $\|\cdot\|$ be defined by (9.2). Then for all $u, v \in V$

$$|\langle u, v \rangle| \leq \|u\| \cdot \|v\|.$$

Moreover the equality holds iff u and v are linearly dependent.

Proof. Let $\|v\| \neq 0$. Take $\lambda = \frac{\langle u, v \rangle}{\|v\|^2}$

We compute

$$\begin{aligned} 0 \leq \|u - \lambda v\|^2 &= \langle u - \lambda v, u - \lambda v \rangle = \langle u, u \rangle - \\ &- \langle \lambda v, u \rangle - \langle u, \lambda v \rangle + \langle \lambda v, \lambda v \rangle = \langle u, u \rangle - \\ &- \lambda \overline{\langle u, v \rangle} - \bar{\lambda} \langle u, v \rangle + \lambda \bar{\lambda} \langle v, v \rangle = \\ &= \|u\|^2 - \frac{|\langle u, v \rangle|^2}{\|v\|^2} - \frac{|\langle u, v \rangle|^2}{\|v\|^2} + \frac{|\langle u, v \rangle|^2}{\|v\|^2} = \\ &= \|u\|^2 - \frac{|\langle u, v \rangle|^2}{\|v\|^2}. \quad \square \end{aligned}$$