

Lecture 25

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April 7, 2003

1 Euclidean spaces

Definition 1.1. Let V be a vector space. Suppose to any 2 vectors $v, u \in V$ there assigned a number from \mathbb{R} which will be denoted by $\langle v, u \rangle$ such that the following 3 properties hold:

Bilinearity • $\langle au_1 + bu_2, v \rangle = a\langle u_1, v \rangle + b\langle u_2, v \rangle$

• $\langle u, av_1 + bv_2 \rangle = a\langle u, v_1 \rangle + b\langle u, v_2 \rangle$

Reflexivity $\langle u, v \rangle = \langle v, u \rangle$

Positivity $\langle u, u \rangle \geq 0$; moreover if $\langle u, u \rangle = 0$, then $u = 0$.

Any function which satisfy properties above is called a **scalar (inner) product**. A vector space V with a scalar product is called a **(real) Euclidean space**.

Now we will give popular examples of the scalar products in different spaces.

\mathbb{R}^n The scalar product of 2 vectors x and y from \mathbb{R}^n can be defined as following: if

$$x = (x_1, x_2, \dots, x_n) \text{ and } y = (y_1, y_2, \dots, y_n)$$

then

$$\langle x, y \rangle = x_1y_1 + x_2y_2 + \dots + x_ny_n.$$

If the vectors are represented by column-vectors, i.e. by $n \times 1$ -matrices, then

$$\langle x, y \rangle = x^\top y.$$

Example 1.2. Let $x = (1, 2, 3)$ and $y = (3, -1, 4)$. Then $\langle x, y \rangle = 1 \cdot 3 + 2 \cdot (-1) + 3 \cdot 4 = 13$.

Actually, there are other ways of defining a scalar product in the vector space \mathbb{R}^n . For example, given positive numbers a_1, a_2, \dots, a_n we can define the scalar product to be

$$\langle (x_1, x_2, \dots, x_n), (y_1, y_2, \dots, y_n) \rangle = a_1(x_1y_1) + a_2(x_2y_2) + \dots + a_n(x_ny_n).$$

$M_{m,n}$ The scalar product of 2 $m \times n$ -matrices A and B such that

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \dots & \dots & \dots & \dots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \text{ and } B = \begin{pmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \dots & \dots & \dots & \dots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{pmatrix}$$

is equal to the sum of products of the corresponding entries:

$$\begin{aligned} \langle A, B \rangle &= a_{11}b_{11} + a_{12}b_{12} + \dots + a_{1n}b_{1n} \\ &+ a_{21}b_{21} + a_{22}b_{22} + \dots + a_{2n}b_{2n} \\ &+ \dots \\ &+ a_{m1}b_{m1} + a_{m2}b_{m2} + \dots + a_{mn}b_{mn}. \end{aligned}$$

We can write this scalar product in terms of matrix multiplication, but to do it we need one more definition.

Definition 1.3. The **trace** of the square matrix A is the sum of its diagonal elements. It is denoted by $\text{tr } A$.

Example 1.4.

$$\text{tr} \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} = 1 + 5 + 9 = 15.$$

Now, using trace we can write that

$$\langle A, B \rangle = \text{tr}(AB^T).$$

This is true, since diagonal elements of AB^T are the following:

$$\begin{aligned} (1, 1) : & a_{11}b_{11} + a_{12}b_{12} + \dots + a_{1n}b_{1n}; \\ (2, 2) : & a_{21}b_{21} + a_{22}b_{22} + \dots + a_{2n}b_{2n}; \\ & \dots \\ (m, m) : & a_{m1}b_{m1} + a_{m2}b_{m2} + \dots + a_{mn}b_{mn}. \end{aligned}$$

$P_n(t)$, $P(t)$, $C[0, 1]$ Here we're considering the spaces of polynomials and the space of continuous functions on the interval $[0, 1]$. If f, g are 2 functions (or polynomials) we can define their scalar product by the following formula:

$$\langle f, g \rangle = \int_0^1 f(t)g(t) dt.$$

Moreover, we can consider the space of all functions which are continuous on the interval $[a, b]$, and in this case the scalar product will be defined as

$$\langle f, g \rangle = \int_a^b f(t)g(t) dt.$$

In all mentioned cases it is simple to check that the properties of the scalar product are satisfied.

2 Norm. Cauchy-Bunyakovsky-Schwartz inequality

In the vector space we can define the function, which is the natural generalization of the length of the vector.

Definition 2.1. For any vector v from the Euclidean space we can define the number $\|v\|$, which is called the **norm** of v by the following formula:

$$\|v\| = \sqrt{\langle v, v \rangle}$$

In the case of the space \mathbb{R}^n we see that the norm of a vector is its length:

$$\|(x_1, x_2, \dots, x_n)\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

The following important fact about the scalar product is one of the most fundamental theorems in mathematics.

Theorem 2.2 (Cauchy-Bunyakovsky-Schwartz inequality). For any vectors u and v from the Euclidean space V ,

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

In the different form this inequality can be written as

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

Proof. Let t be an arbitrary number. Let's consider the scalar product of $tu + v$ with itself:

$$\begin{aligned} \langle tu + v, tu + v \rangle &= t^2 \langle u, u \rangle + t \langle u, v \rangle + t \langle v, u \rangle + \langle v, v \rangle \\ &= t^2 \langle u, u \rangle + 2t \langle u, v \rangle + \langle v, v \rangle \end{aligned}$$

This can be considered as a polynomial of the second power of t . Since $\langle tu + v, tu + v \rangle \geq 0$, then this quadratic polynomial is greater or equal to 0. So, it has either 1 or no roots, and thus its discriminant is less than or equal to 0:

$$D = \langle u, v \rangle^2 - \langle u, u \rangle \langle v, v \rangle \leq 0.$$

Thus,

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

□

Using this theorem we can now consider the properties of the norm.

Positivity $\|v\| \geq 0$. Moreover, $\|v\| = 0$ if and only if $v = \mathbf{0}$.

Linearity $\|kv\| = |k|\|v\|$.

Triangle inequality $\|u + v\| \leq \|u\| + \|v\|$.

Proof. Positivity: Directly follows from the definition of the norm.

Linearity: $\|kv\| = \sqrt{\langle kv, kv \rangle} = \sqrt{k^2 \langle v, v \rangle} = |k| \sqrt{\langle v, v \rangle} = |k| \|v\|$.

Triangle inequality: Let's consider $\|u + v\| \geq 0$. We can rewrite it in the following way:

$$\begin{aligned} \|u + v\|^2 &= \langle u + v, u + v \rangle \\ &= \langle u, u \rangle + 2\langle u, v \rangle + \langle v, v \rangle \\ &= \|u\|^2 + 2\langle u, v \rangle + \|v\|^2 && \text{by C-B-S inequality} \\ &\leq \|u\|^2 + 2\|u\|\|v\| + \|v\|^2 \\ &= (\|u\| + \|v\|)^2 \end{aligned}$$

Now, taking roots of both sides we get

$$\|u + v\| \leq \|u\| + \|v\|.$$

□

Geometrically speaking, the last inequality means that the side of the triangle is less than or equal to the sum of other 2 sides:

