

Module MAU23203: Analysis in Several Real
Variables

Michaelmas Term 2023

Section 2: Schwarz's Inequality and some
Related Inequalities

D. R. Wilkins

Copyright © Trinity College Dublin 2020–22

Contents

2	Schwarz's Inequality and some Related Inequalities	5
2.1	Basic Properties of Vectors and Norms	5
2.2	The Hilbert-Schmidt Norm of a Linear Transformation	6

2 Schwarz's Inequality and some Related Inequalities

2.1 Basic Properties of Vectors and Norms

We come now to a discussion of inequalities satisfied by vectors in the vector space \mathbb{R}^n of dimension n whose elements are ordered n -tuples of real numbers. The most basic inequality we consider here is *Schwarz's Inequality*. This inequality is then applied in the proof of the *Triangle Inequality*. It is further applied in the proof of some inequalities involving linear transformations between finite-dimensional vector spaces.

The set \mathbb{R}^n of ordered n -tuples of real numbers represents n -dimensional *Euclidean space* (with respect to the standard Cartesian coordinate system).

Let \mathbf{x} and \mathbf{y} be elements of \mathbb{R}^n , where

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

and let c be a real number. We define

$$\begin{aligned} \mathbf{x} + \mathbf{y} &= (x_1 + y_1, x_2 + y_2, \dots, x_n + y_n), \\ \mathbf{x} - \mathbf{y} &= (x_1 - y_1, x_2 - y_2, \dots, x_n - y_n), \\ c\mathbf{x} &= (cx_1, cx_2, \dots, cx_n), \\ \mathbf{x} \cdot \mathbf{y} &= x_1y_1 + x_2y_2 + \dots + x_ny_n, \\ |\mathbf{x}| &= \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}. \end{aligned}$$

The quantity $\mathbf{x} \cdot \mathbf{y}$ is the *scalar product* (or *inner product*) of \mathbf{x} and \mathbf{y} , and the quantity $|\mathbf{x}|$ is the *Euclidean norm* of \mathbf{x} . Note that $|\mathbf{x}|^2 = \mathbf{x} \cdot \mathbf{x}$. The *Euclidean distance* between two points \mathbf{x} and \mathbf{y} of \mathbb{R}^n is defined to be the Euclidean norm $|\mathbf{y} - \mathbf{x}|$ of the vector $\mathbf{y} - \mathbf{x}$.

Proposition 2.1 (Schwarz's Inequality) *Let \mathbf{x} and \mathbf{y} be elements of \mathbb{R}^n . Then $|\mathbf{x} \cdot \mathbf{y}| \leq |\mathbf{x}||\mathbf{y}|$.*

Proof We note that $|t\mathbf{x} + \mathbf{y}|^2 \geq 0$ for all real numbers t . But

$$|t\mathbf{x} + \mathbf{y}|^2 = (t\mathbf{x} + \mathbf{y}) \cdot (t\mathbf{x} + \mathbf{y}) = t^2|\mathbf{x}|^2 + 2t\mathbf{x} \cdot \mathbf{y} + |\mathbf{y}|^2.$$

Therefore $t^2|\mathbf{x}|^2 + 2t\mathbf{x} \cdot \mathbf{y} + |\mathbf{y}|^2 \geq 0$ for all real numbers t . Thus $at^2 + bt + c \geq 0$ for all real numbers t , where $a = |\mathbf{x}|^2$, $b = 2\mathbf{x} \cdot \mathbf{y}$ and $c = |\mathbf{y}|^2$.

Now $at^2 + bt + c$ is a quadratic polynomial in the real variable t whose values must be non-negative for all real values of t . A necessary and sufficient

condition for this to be the case is that the inequality $b^2 \leq 4ac$ be satisfied. Thus, substituting in the values for a , b and c previously given, we find that

$$(\mathbf{x} \cdot \mathbf{y})^2 \leq |\mathbf{x}|^2 |\mathbf{y}|^2.$$

Schwarz's inequality now follows on taking the positive square roots of both sides. ■

Proposition 2.2 (Triangle Inequality) *Let \mathbf{x} and \mathbf{y} be elements of \mathbb{R}^n . Then $|\mathbf{x} + \mathbf{y}| \leq |\mathbf{x}| + |\mathbf{y}|$.*

Proof Using Schwarz's Inequality, we see that

$$\begin{aligned} |\mathbf{x} + \mathbf{y}|^2 &= (\mathbf{x} + \mathbf{y}) \cdot (\mathbf{x} + \mathbf{y}) = |\mathbf{x}|^2 + |\mathbf{y}|^2 + 2\mathbf{x} \cdot \mathbf{y} \\ &\leq |\mathbf{x}|^2 + |\mathbf{y}|^2 + 2|\mathbf{x}||\mathbf{y}| = (|\mathbf{x}| + |\mathbf{y}|)^2. \end{aligned}$$

The result follows directly. ■

It follows immediately from the Triangle Inequality (Proposition 2.2) that

$$|\mathbf{z} - \mathbf{x}| \leq |\mathbf{z} - \mathbf{y}| + |\mathbf{y} - \mathbf{x}|$$

for all points \mathbf{x} , \mathbf{y} and \mathbf{z} of \mathbb{R}^n . This important inequality expresses the geometric fact that the length of any one side of a triangle in a Euclidean space is less than or equal to the sum of the lengths of the other two sides of that triangle.

2.2 The Hilbert-Schmidt Norm of a Linear Transformation

Recall that the *length* (or *norm*) of an element $\mathbf{x} \in \mathbb{R}^n$ is defined such that

$$|\mathbf{x}|^2 = x_1^2 + x_2^2 + \cdots + x_n^2.$$

Definition Let $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$ be a linear transformation from \mathbb{R}^m to \mathbb{R}^n , and let $(T_{i,j})$ be the $n \times m$ matrix representing this linear transformation with respect to the standard bases of \mathbb{R}^m and \mathbb{R}^n . The *Hilbert-Schmidt norm* $\|T\|_{\text{HS}}$ of the linear transformation is then defined so that

$$\|T\|_{\text{HS}} = \sqrt{\sum_{i=1}^n \sum_{j=1}^m T_{i,j}^2}.$$

Note that the Hilbert-Schmidt norm is just the Euclidean norm on the real vector space of dimension mn whose elements are $n \times m$ matrices representing linear transformations from \mathbb{R}^m to \mathbb{R}^n with respect to the standard bases of these vector spaces. Therefore it has the standard properties of the Euclidean norm. In particular it follows from the Triangle Inequality (Lemma 2.2) that

$$\|T + U\|_{\text{HS}} \leq \|T\|_{\text{HS}} + \|U\|_{\text{HS}} \quad \text{and} \quad \|cT\|_{\text{HS}} = |c| \|T\|_{\text{HS}}$$

for all linear transformations T and U from \mathbb{R}^m to \mathbb{R}^n and for all real numbers c .

Proposition 2.3 *Let $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$ be a linear transformation from \mathbb{R}^m to \mathbb{R}^n . Then*

$$|T\mathbf{x}| \leq \|T\|_{\text{HS}} |\mathbf{x}|$$

for all $\mathbf{x} \in \mathbb{R}^m$, where $\|T\|_{\text{HS}}$ is the Hilbert-Schmidt norm of the linear transformation T .

Proof Let $\mathbf{v} = T\mathbf{x}$, and let $\mathbf{v} = (v_1, v_2, \dots, v_n)$. Then

$$v_i = T_{i,1}x_1 + T_{i,2}x_2 + \dots + T_{i,m}x_m$$

for all integers i between 1 and n . It follows from Schwarz's Inequality (Lemma 2.1) that

$$v_i^2 \leq \left(\sum_{j=1}^m T_{i,j}^2 \right) \left(\sum_{j=1}^m x_j^2 \right) = \left(\sum_{j=1}^m T_{i,j}^2 \right) |\mathbf{x}|^2.$$

Hence

$$|\mathbf{v}|^2 = \sum_{i=1}^n v_i^2 \leq \left(\sum_{i=1}^n \sum_{j=1}^m T_{i,j}^2 \right) |\mathbf{x}|^2 = \|T\|_{\text{HS}}^2 |\mathbf{x}|^2.$$

Thus $|T\mathbf{x}| \leq \|T\|_{\text{HS}} |\mathbf{x}|$, as required. ■

The following corollary follows immediately from Proposition 2.3.

Corollary 2.4 *Let $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$ be a linear transformation from \mathbb{R}^m to \mathbb{R}^n . Then there exists a positive real number K , dependent on the choice of linear transformation T but independent of \mathbf{x} , with the property that*

$$|T\mathbf{x}| \leq K |\mathbf{x}|$$

for all $\mathbf{x} \in \mathbb{R}^m$.

In certain proofs in real analysis involving linear transformations, it is sufficient for the purposes of the proof that there should exist some positive constant K for which the inequality in the statement of Corollary 2.4 is satisfied. Nevertheless, if a more precise estimate of the value of such a constant K were required, then Proposition 2.3 would provide more information.

Lemma 2.5 *Let $T: \mathbb{R}^m \rightarrow \mathbb{R}^n$ be a linear transformation from \mathbb{R}^m to \mathbb{R}^n and let $S: \mathbb{R}^n \rightarrow \mathbb{R}^p$ be a linear transformation from \mathbb{R}^n to \mathbb{R}^p . Then the Hilbert-Schmidt norm of the composition of the linear transformations T and S satisfies the inequality $\|ST\|_{\text{HS}} \leq \|S\|_{\text{HS}} \|T\|_{\text{HS}}$.*

Proof The composition ST of the linear transformations is represented by the product of the corresponding matrices. Thus the component $(ST)_{k,j}$ in the k th row and the j th column of the $p \times m$ matrix representing the linear transformation ST satisfies

$$(ST)_{k,j} = \sum_{i=1}^n S_{k,i} T_{i,j}.$$

where $S_{k,i}$ and $T_{i,j}$ denote the components in the relevant rows and columns of the matrices representing the linear transformations S and T respectively. It follows from Schwarz's Inequality (Lemma 2.1) that

$$(ST)_{k,j}^2 \leq \left(\sum_{i=1}^n S_{k,i}^2 \right) \left(\sum_{i=1}^n T_{i,j}^2 \right).$$

Summing over k , we find that

$$\sum_{k=1}^p (ST)_{k,j}^2 \leq \left(\sum_{k=1}^p \sum_{i=1}^n S_{k,i}^2 \right) \left(\sum_{i=1}^n T_{i,j}^2 \right) = \|S\|_{\text{HS}}^2 \left(\sum_{i=1}^n T_{i,j}^2 \right).$$

Then summing over j , we find that

$$\begin{aligned} \|ST\|_{\text{HS}}^2 &= \sum_{k=1}^p \sum_{j=1}^m (ST)_{k,j}^2 \leq \|S\|_{\text{HS}}^2 \left(\sum_{i=1}^n \sum_{j=1}^m T_{i,j}^2 \right) \\ &\leq \|S\|_{\text{HS}}^2 \|T\|_{\text{HS}}^2. \end{aligned}$$

On taking square roots, we find that $\|ST\|_{\text{HS}} \leq \|S\|_{\text{HS}} \|T\|_{\text{HS}}$, as required. ■