

A BRIEF INTRODUCTION TO THE CAUCHY-SCHWARZ AND HÖLDER INEQUALITIES

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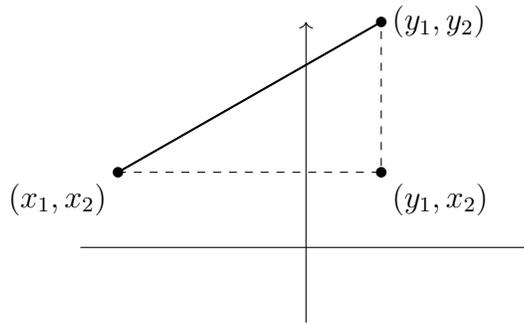
ABSTRACT. The Cauchy-Schwarz inequality is fundamental to many areas of mathematics, physics, engineering, and computer science. We introduce and motivate this inequality, show some applications, and indicate some generalizations, including a simpler form of Hölder’s inequality than is usually presented.

1. MOTIVATING CAUCHY-SCHWARZ

Recall that the standard distance (aka “Euclidean metric”) between two points $\mathbf{x}, \mathbf{y} \in \mathbb{R}^2$ is defined

$$d(\mathbf{x}, \mathbf{y}) := \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2},$$

a simple consequence of the pythagorean theorem.



Calculating the distance between $\mathbf{x} = (x_1, x_2)$ and $\mathbf{y} = (y_1, y_2)$

The Euclidean metric famously satisfies the triangle inequality

$$d(\mathbf{x}, \mathbf{y}) \leq d(\mathbf{x}, \mathbf{z}) + d(\mathbf{z}, \mathbf{y}), \tag{1.1}$$

which asserts that the shortest path from one point of the plane to another is via a straight line. This is intuitively clear, but the proof isn’t immediately obvious. We warm up by thinking about a special case:

Proposition 1.1. $d(\mathbf{x}, \mathbf{y}) \leq d(\mathbf{x}, 0) + d(0, \mathbf{y})$ for all $\mathbf{x}, \mathbf{y} \in \mathbb{R}^2$.

Exercise 1. Show that the proposition implies (1.1).

Scratchwork for proof of the Proposition:

Writing out the claim in terms of the formal definition of the Euclidean distance, we see that we’re trying to prove

$$\sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2} \leq \sqrt{x_1^2 + x_2^2} + \sqrt{y_1^2 + y_2^2}.$$

Squaring both sides and simplifying shows this is equivalent to proving

$$-2x_1y_1 - 2x_2y_2 \leq 2\sqrt{x_1^2 + x_2^2}\sqrt{y_1^2 + y_2^2}.$$

Dividing by 2 and squaring both sides, we see that our claim is equivalent to

$$(x_1y_1 + x_2y_2)^2 \leq (x_1^2 + x_2^2)(y_1^2 + y_2^2). \tag{†}$$

Thus the triangle inequality for the Euclidean metric in \mathbb{R}^2 is equivalent to the inequality (†). A similar argument shows that the triangle inequality for the Euclidean metric in \mathbb{R}^n is equivalent to the following:

Theorem 1.2 (Cauchy-Schwarz). For any real numbers x_i, y_i we have

$$(x_1y_1 + x_2y_2 + \cdots + x_ny_n)^2 \leq (x_1^2 + x_2^2 + \cdots + x_n^2)(y_1^2 + y_2^2 + \cdots + y_n^2).$$

We'll prove this below, but for now let's explore some easy consequences of this inequality. Taking all the $y_i = 1$, we immediately deduce

$$(x_1 + x_2 + \cdots + x_n)^2 \leq (x_1^2 + x_2^2 + \cdots + x_n^2)n. \quad (*)$$

Rearranging yields

$$\frac{x_1 + x_2 + \cdots + x_n}{n} \leq \sqrt{\frac{x_1^2 + x_2^2 + \cdots + x_n^2}{n}},$$

or in words, that the mean is bounded above by the root mean square. This already hints at a connection to probability and statistics. Inspired by this, let's impose the additional condition $x_1 + x_2 + \cdots + x_n = 1$ (the sum of probabilities of all possible events is 1); if this is the case, (*) implies

$$x_1^2 + x_2^2 + \cdots + x_n^2 \geq \frac{1}{n}.$$

For example, this shows that if I roll a 6-sided die twice, the probability that both rolls produce the same number is at least $1/6$ – *no matter whether or not the die is fairly weighted*.

Let's reconsider the original Cauchy-Schwarz inequality from a different perspective. What does the quantity $x_1y_1 + x_2y_2 + \cdots + x_ny_n$ remind you of? The dot product of $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$! Thus we can rewrite Cauchy-Schwarz in the more compact form

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

This change of perspective is not merely notationally convenient, but also suggests a short proof. Recall that for any two vectors, $\mathbf{a} \cdot \mathbf{b} = |\mathbf{a}| |\mathbf{b}| \cos \theta$ where θ is the angle between \mathbf{a} and \mathbf{b} (this is a consequence of the law of cosines).

Proof of Cauchy-Schwarz. Given $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, we have

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

In fact, examining this proof we see that equality holds in Cauchy-Schwarz iff the angle between \mathbf{x} and \mathbf{y} is a multiple of π , or in other words, iff \mathbf{x} is a rescaling of \mathbf{y} . Thus, we can write the theorem in a stronger form:

Theorem 1.3 (Cauchy-Schwarz, v2.0). Given $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, we have

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

with equality if and only if \mathbf{x} is a rescaling of \mathbf{y} .

This refinement of Cauchy-Schwarz is quite useful, as the following example demonstrates.

Example 1. What is the maximum of $2x + 3y + 5z$ over all the points (x, y, z) on the unit sphere?

By Cauchy-Schwarz we have

$$(2x + 3y + 5z)^2 \leq (2^2 + 3^2 + 5^2)(x^2 + y^2 + z^2) = 38,$$

with equality iff (x, y, z) is a rescaling of $(2, 3, 5)$. Thus the function $2x + 3y + 5z$ takes on a maximal value of $\sqrt{38}$, and this happens precisely at the points $\pm \left(\frac{2}{\sqrt{38}}, \frac{3}{\sqrt{38}}, \frac{5}{\sqrt{38}} \right)$.

We've now seen applications of Cauchy-Schwarz to problems in geometry, probability, and optimization. This is the tip of the iceberg; Cauchy-Schwarz is extremely useful throughout mathematics, physics, engineering, and computer science.

2. GENERALIZING CAUCHY-SCHWARZ

Returning to Cauchy-Schwarz in the form presented in Theorem 1.2, observe that the right hand side of the inequality doesn't change if we replace any of the x_i by $-x_i$. Right away we deduce the stronger bound

$$(|x_1y_1| + |x_2y_2| + \cdots + |x_ny_n|)^2 \leq (x_1^2 + x_2^2 + \cdots + x_n^2)(y_1^2 + y_2^2 + \cdots + y_n^2).$$

The take-home message here is that it suffices to state the Cauchy-Schwarz inequality (as well as other inequalities) for the special case of *non-negative* real numbers; inserting absolute values often produces a stronger version for free!

As a student I found Cauchy-Schwarz difficult to remember, and came up with the following mnemonic:

The square of the sum of products \leq the product of the sum of squares.

To be useful one must remember the direction of the inequality, but this doesn't need to be memorized: a single random example suffices to determine this, e.g. $\mathbf{x} = (1, 0)$ and $\mathbf{y} = (0, 1)$. As with any mnemonic, however, there's a possibility of misinterpretation. In this case, the word *product* is potentially misleading; how many variables are we taking the product of? This is important, since taking the trivial product of a single variable leads to the false conclusion $(a + b)^2 \leq a^2 + b^2$. What we actually mean by "product" is the product of *two* variables. Thus a more precise version of the mnemonic might read *the square of the sum of products (of two variables) \leq the product of the sum of the squares*, but this is clunkier so in practice I suppress the parenthetical remark.

One nice feature of the above mnemonic is that it inspires generalizations. For example, one might guess

The cube of the sum of products (of three variables) \leq the product of the sum of cubes.

This turns out to be true, with one caveat: all the variables have to be non-negative. (As above, one can remove this restriction by inserting absolute values into the inequality.) Taking this idea and running with it, we might be led to conjecture the following:

Theorem 2.1 (Hölder's inequality). *For any positive integer m , we have*

$$\left(\sum_i \underbrace{\alpha_i \beta_i \cdots \omega_i}_{m \text{ variables}} \right)^m \leq \underbrace{\left(\sum_i \alpha_i^m \right) \left(\sum_i \beta_i^m \right) \cdots \left(\sum_i \omega_i^m \right)}_{m \text{ factors}} \quad (\heartsuit)$$

assuming all of the $\alpha_i, \beta_i, \dots, \omega_i$ are non-negative real numbers.

The proof of this is outlined in the exercises. Just as Cauchy-Schwarz is the natural tool for proving the triangle inequality in \mathbb{R}^n with respect to the Euclidean metric, Hölder's inequality is useful for proving the triangle inequality in some other spaces that arise in analysis (called L^p spaces).

Recasting Cauchy-Schwarz in the language of vectors (as in Theorem 1.3) provides a different avenue for generalization, to arbitrary vector spaces endowed with an *inner product* (a generalization of the dot product). We won't discuss this here, but you will encounter this in any course on functional analysis or Hilbert spaces. An example of this is a version of Cauchy-Schwarz for integrals rather than sums; see exercise 1.d below.

3. EXERCISES

1. Practice with Cauchy-Schwarz:

- (a) Prove that $|x_1y_1 + \cdots + x_ny_n|^2 \leq (|x_1|^2 + \cdots + |x_n|^2)(|y_1|^2 + \cdots + |y_n|^2)$ for any $x_i, y_i \in \mathbb{C}$.
- (b) Prove that for any $a, b, c > 0$,

$$\frac{a}{b+c} + \frac{b}{a+c} + \frac{c}{a+b} \geq \frac{3}{2}.$$

- (c) Given $\theta_1, \theta_2, \dots, \theta_n \in [0, \frac{\pi}{2})$, prove that

$$\frac{1}{\cos^2 \theta_1} + \cdots + \frac{1}{\cos^2 \theta_n} \geq n + \frac{(\sin \theta_1 + \cdots + \sin \theta_n)^2}{\cos^2 \theta_1 + \cdots + \cos^2 \theta_n}.$$

- (d) Prove that $\left| \int_a^b f(x)g(x) dx \right|^2 \leq \left(\int_a^b |f(x)|^2 dx \right) \left(\int_a^b |g(x)|^2 dx \right)$.
2. The purpose of this problem is to discover a few more proofs of Cauchy-Schwarz.
- (a) Prove directly that $(x_1y_1 + x_2y_2)^2 \leq (x_1^2 + x_2^2)(y_1^2 + y_2^2)$. Then deduce Theorem 1.2 by induction.
- (b) Given real numbers a_i and b_i , consider the function

$$F(x) := \sum_i (a_i x - b_i)^2.$$

Observe that $F(x)$ is quadratic in x , i.e. can be written in the form $Ax^2 + Bx + C$. What can you deduce about the discriminant $B^2 - 4AC$? Use this to prove Theorem 1.3.

- (c) Prove that

$$\left(\sum_i x_i y_i \right)^2 = \left(\sum_i x_i^2 \right) \left(\sum_i y_i^2 \right) - \frac{1}{2} \sum_{i,j} (x_i y_j - x_j y_i)^2,$$

and deduce Theorem 1.3.

3. The goal of this problem is to prove Theorem 2.1, i.e. that the inequality (\heartsuit) holds for all positive integers m .
- (a) Prove that if (\heartsuit) holds when $m = a$ and when $m = b$, then it also must hold for $m = ab$.
- (b) Prove that (\heartsuit) holds whenever m is a power of 2.
- (c) Prove that for any non-negative real numbers a_n, b_n, c_n we have

$$\sum a_n b_n c_n \leq \left(\sum a_n b_n c_n \right)^{1/4} \left(\sum a_n^3 \right)^{1/4} \left(\sum b_n^3 \right)^{1/4} \left(\sum c_n^3 \right)^{1/4}.$$

- (d) Prove (\heartsuit) for $m = 3$. [Hint: use the previous part!]

- (e) Prove Theorem 2.1. [Hint: use strong induction. m is either even or odd...]

4. Frequently in the literature, Hölder's inequality refers to the bound

$$a_1 b_1 + \cdots + a_n b_n \leq (a_1^p + \cdots + a_n^p)^{1/p} (b_1^q + \cdots + b_n^q)^{1/q} \quad (\ddagger)$$

for any positive real numbers p, q satisfying $1/p + 1/q = 1$ and any non-negative $a_i, b_i \in \mathbb{R}$. The goal of this exercise is to prove that (\ddagger) is equivalent to Theorem 2.1.

- (a) Prove that (\ddagger) implies Theorem 2.1.

- (b) Show that Theorem 2.1 implies (\ddagger) for all $p \in \mathbb{Q}$. [Hint: Warm up with the case $p = 3$.]

- (c) Deduce that (\ddagger) holds for all real numbers $p > 1$. [Hint: Think of the numbers a_i, b_i being fixed and of p as the variable.]

5. In this exercise we outline a beautiful argument (due to Orr Shalit) that deduces (\ddagger) directly from Cauchy-Schwarz.¹ Throughout, let

$$S := \{1/p \in (0, 1) : \text{the bound } (\ddagger) \text{ holds for all positive real choices of } a_i, b_i\}.$$

- (a) Prove that if $\frac{1}{p} \in S$, then $\frac{1}{2p} \in S$. [Hint: Write $a_k b_k = (a_k b_k^\sigma)(b_k^{1-\sigma})$, and apply Cauchy-Schwarz to $\sum a_k b_k$. Then choose σ appropriately.]

- (b) Deduce that $\frac{a}{2^k} \in S$ for all positive integers k and all $a \in \{1, 2, \dots, 2^k - 1\}$. [Hint: Do you happen to know any elements of S a priori?]

- (c) Conclude that $S = (0, 1)$.

¹For a completely different proof of (\ddagger), see my article *Hölder's inequality as a convexity result*, available online.