

3/5/2019: Norms.

The absolute value function is a special example of a norm on the vector space of real numbers,  $V = \mathbb{R}$ .

$$|\cdot| : \mathbb{R} \rightarrow \mathbb{R}^+ \cup \{0\} \text{ ~~satisfying~~}$$

$$\text{defined as: } |v| = \begin{cases} v & \text{if } v \geq 0. \\ -v & \text{if } v < 0. \end{cases}$$

Properties:

$$\rightarrow |v| \geq 0 \quad \forall v \in \mathbb{R}, \quad |v| = 0 \Leftrightarrow v = 0.$$

$$\rightarrow |\lambda v| = |\lambda| |v| \quad \forall \lambda \in \mathbb{R} \text{ and } \forall v \in \mathbb{R}$$

$$\rightarrow |u+v| \leq |u| + |v| \quad \forall u, v \in \mathbb{R}.$$

We used this norm to measure convergence of sequences. For example, we know the sequence of iterates from Newton's method satisfies:

$$|x_{k+1} - \xi| \leq C |x_k - \xi|^2.$$

i.e. a norm gives us a measure

of distance in a vector space  $V$ .

Definition 2.6 Suppose  $V$  is a vector space. (What does this mean?).  $V$  is closed under addition and scalar multiplication. Here, the scalars are just real numbers. We say  $\|\cdot\|$  is a norm on  $V$  if it satisfies:

- ①  $\|\underline{v}\| = 0 \iff \underline{v} = 0$  in  $V$
- ②  $\|\lambda \underline{v}\| = |\lambda| \|\underline{v}\| \quad \forall \lambda \in \mathbb{R} \text{ and } \forall \underline{v} \in V$
- ③  $\|\underline{u} + \underline{v}\| \leq \|\underline{u}\| + \|\underline{v}\| \quad \forall \underline{u}, \underline{v} \in V.$

The ~~map~~ pair  $(V, \|\cdot\|)$  is called a normed linear space, or a normed vector space.

Some examples of norms on the vector space  $V = \mathbb{R}^n$ , given as definitions since these norms are important.

Definition 2.7: "1-norm"

$$\|\underline{v}\|_1 = \sum_{i=1}^n |v_i|.$$

Definition 2.8 "2-norm"

$$\|\underline{v}\|_2 = \left( \sum_{i=1}^n v_i^2 \right)^{1/2}$$

Note that you can write the 2-norm as

$$(\underline{v} \cdot \underline{v})^{1/2} = \|\underline{v}\|_2.$$

i.e. this norm is "induced" from an inner product on  $\mathbb{R}^n$ .

Definition 2.9 The " $\infty$ -norm" is

$$\|\underline{v}\|_\infty = \max_{i=1, \dots, n} |v_i|.$$

Remark: when  $n=1$ , i.e.  $V = \mathbb{R}$ , then

$$\|\underline{v}\|_1 = \|\underline{v}\|_2 = \|\underline{v}\|_\infty = |\underline{v}| = |v|.$$

Example: let  $V = \mathbb{R}^3$ ,  $\underline{v} = (1, -2, 3.14)$ .

$$\|\underline{v}\|_1 = 1 + 2 + 3.14 = 6.14.$$

$$\|\underline{v}\|_2 = \sqrt{1 + 4 + 3.14^2} = 3.8596$$

$$\|\underline{v}\|_\infty = \max\{|1|, |-2|, |3.14|\} = 3.14.$$

Very important:

Cauchy-Schwarz Inequality:  $\left| \sum_{i=1}^n u_i v_i \right| \leq \|\underline{u}\|_2 \|\underline{v}\|_2$   
 $\forall \underline{u}, \underline{v} \in \mathbb{R}^n$

Cauchy-Schwarz is a tool useful for many many things. In particular, we will use it to show the triangle inequality.

### Proof of Cauchy-Schwarz

Take  $\lambda \in \mathbb{R}$ ,  $u, v \in \mathbb{R}^n$  arbitrary.

$$\begin{aligned} 0 &\leq \| \lambda u + v \|_2^2 = \sum_{i=1}^n (\lambda u_i + v_i)^2 \\ &= \underbrace{\lambda^2 \sum_{i=1}^n |u_i|^2}_{= A} + 2\lambda \underbrace{\sum_{i=1}^n u_i v_i}_{= B} + \underbrace{\sum_{i=1}^n |v_i|^2}_{= C}. \end{aligned}$$

We have a nonnegative quadratic polynomial in  $\lambda$ :

$$f(\lambda) = A \lambda^2 + B \lambda + C \geq 0.$$

So, either  $f(\lambda)$  has only one root:

$$B^2 - 4AC = 0$$

or it has no real roots, i.e.

$$B^2 - 4AC < 0.$$

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So we have in either case.

$$B^2 - 4AC \leq 0$$

$$\Leftrightarrow \left( 2 \sum_{i=1}^n u_i v_i \right)^2 - 4 \left( \sum_{i=1}^n u_i^2 \right) \left( \sum_{i=1}^n v_i^2 \right) \leq 0$$

$$\Rightarrow \left( \sum_{i=1}^n u_i v_i \right)^2 \leq \left( \sum_{i=1}^n u_i^2 \right) \left( \sum_{i=1}^n v_i^2 \right).$$

Taking the square root of both sides gives:

$$\left( \sum_{i=1}^n u_i v_i \right) \leq \| \underline{u} \|_2 \| \underline{v} \|_2.$$

Remark: Equality holds in Cauchy-Schwarz

if and only if  $(\Leftrightarrow) \underline{u} = \lambda \underline{v}$  for some  $\lambda \in \mathbb{R}$ ,  $\lambda \geq 0$ .

$(\Leftarrow)$  Easier direction: let  $\underline{u} = \lambda \underline{v}$ .

$$\begin{aligned} \text{Then } \| \underline{u} \| \| \underline{v} \| &= \| \lambda \underline{v} \| \| \underline{v} \| = |\lambda| \| \underline{v} \|^2 \\ &= |\lambda| \underline{v} \cdot \underline{v} = \lambda \underline{v} \cdot \underline{v} = \underline{u} \cdot \underline{v}. \end{aligned}$$

( $\Rightarrow$ ) Assume equality i.e.  $\underline{u} \cdot \underline{v} = \|\underline{u}\| \|\underline{v}\|$ .

Note that  $\|\underline{u}\| \|\underline{v}\| \cos \theta = \underline{u} \cdot \underline{v}$ ,

with  $\theta =$  angle between vectors  $\underline{u}$  and  $\underline{v}$ .

Assume both  $\underline{u}$  and  $\underline{v}$  are not the zero vector, for if that were true then the result trivially holds. Then

$$\underline{u} \cdot \underline{v} = \|\underline{u}\| \|\underline{v}\| \Leftrightarrow \|\underline{u}\| \|\underline{v}\| \cos \theta$$

$$= \|\underline{u}\| \|\underline{v}\|$$

$$\Leftrightarrow \cos \theta = 1,$$

So  $\theta = 0$ , i.e. vectors  $\underline{u}$  and  $\underline{v}$  are parallel. So  $\underline{u} = \lambda \underline{v}$ , For some  $\lambda \in \mathbb{R}$ ,  $\lambda \geq 0$ .

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Now, we can use Cauchy-Schwarz to prove the triangle inequality for the 2-norm:

$$\|\underline{u} + \underline{v}\|_2^2 = \|\underline{u}\|_2^2 + 2 \sum_{i=1}^n u_i v_i + \|\underline{v}\|_2^2$$

$$\begin{aligned} \left( \text{Cauchy Schwarz} \right) &\leq \|\underline{u}\|_2^2 + 2 \|\underline{u}\|_2 \|\underline{v}\|_2 + \|\underline{v}\|_2^2 \\ &= \left( \|\underline{u}\|_2 + \|\underline{v}\|_2 \right)^2 \end{aligned}$$

implying  $\|\underline{u} + \underline{v}\|_2 \leq \|\underline{u}\|_2 + \|\underline{v}\|_2$ .

Definition: "p-norm." Take  $p \geq 1$ .

it is defined as

$$\|\underline{v}\|_p = \left( \sum_{i=1}^n |v_i|^p \right)^{1/p} \quad \forall \underline{v} \in \mathbb{R}^n$$

How do we show the triangle inequality in general for the p-norm?

In the process of showing that  $\|\cdot\|_p$  satisfies the triangle inequality, we need some tools which are important in their own right.

### Thm 2.4 (Young's Inequality)

let  $p, q > 1$ , and  $\frac{1}{p} + \frac{1}{q} = 1$ .

Then, for any  $a, b \geq 0$ , we have

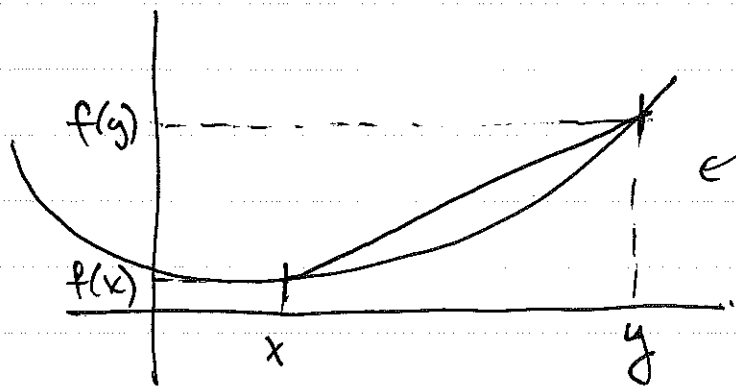
$$ab \leq \frac{a^p}{p} + \frac{b^q}{q}.$$

Proof: Assume  $a > 0$  and  $b > 0$ . If either  $a = 0$  or  $b = 0$ , the result is trivial.

Recall the definition of convexity of a function:  $x \mapsto f(x)$  is convex if

$$f(\theta x + (1-\theta)y) \leq \theta f(x) + (1-\theta)f(y).$$

$$\forall \theta \in [0, 1], \quad \forall x, y \in \mathbb{R}.$$



graph of  $f$   
lies below the  
"chord" connecting  
 $f(x)$  and  $f(y)$ .



Notice that  $x \mapsto e^x$  is convex.  
Doing some manipulations, using the property of logs, we have:

$$\begin{aligned}ab &= e^{\log(ab)} \\ &= e^{\log a + \log b} \\ &= e^{(\frac{1}{p}) \log a^p + (\frac{1}{q}) \log b^q}\end{aligned}$$

Now, apply convexity of  $e^x$  with  $\theta = \frac{1}{p}$ ,  $(1-\theta) = 1 - \frac{1}{p} = \frac{1}{q}$ ,  $x = \log a^p$ ,

$y = \log b^q$ , to conclude

$$\begin{aligned}&e^{(\frac{1}{p}) \log a^p + (\frac{1}{q}) \log b^q} \\ &\leq \frac{1}{p} e^{\log a^p} + \frac{1}{q} e^{\log b^q} \\ &= \frac{a^p}{p} + \frac{b^q}{q},\end{aligned}$$

So we can conclude:

$$ab \leq \frac{a^p}{p} + \frac{b^q}{q}.$$

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Thm: 2.5 (Hölder's inequality).

let  $p, q > 1$ ,  $\frac{1}{p} + \frac{1}{q} = 1$ . Then

$\forall \underline{u}, \underline{v} \in \mathbb{R}^n$  we have:

$$\left| \sum_{i=1}^n u_i v_i \right| \leq \|\underline{u}\|_p \|\underline{v}\|_q.$$

Proof: Assume  $\underline{u} \neq 0$  and  $\underline{v} \neq 0$ , otherwise

the inequality trivially holds. Construct normalized versions of  $\underline{u}$  and  $\underline{v}$ :

$$\underline{\tilde{u}} = \frac{\underline{u}}{\|\underline{u}\|_p}, \quad \underline{\tilde{v}} = \frac{\underline{v}}{\|\underline{v}\|_q}.$$

Now, we look at:

$$\left| \sum_{i=1}^n \tilde{u}_i \tilde{v}_i \right| \leq \sum_{i=1}^n |\tilde{u}_i \tilde{v}_i| \quad (\text{triangle inequality}).$$

$$\begin{aligned} (\text{Young's inequality}) &\leq \frac{1}{p} \sum_{i=1}^n |\tilde{u}_i|^p + \frac{1}{q} \sum_{i=1}^n |\tilde{v}_i|^q \\ &= \frac{1}{p} + \frac{1}{q} = 1. \end{aligned}$$

So, we've shown that

$$\left| \sum_{i=1}^n \tilde{u}_i \tilde{v}_i \right| \leq 1.$$

Plugging in  $\tilde{u}_i = \frac{u_i}{\|u\|_p}$ ,  $\tilde{v}_i = \frac{v_i}{\|v\|_q}$ ,

we conclude  $\left| \sum_{i=1}^n u_i v_i \right| \leq \|u\|_p \|v\|_q$ .

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Thm 2.6: (Minkowski's Inequality).

Let  $1 \leq p \leq \infty$  and  $u, v \in \mathbb{R}^n$ . Then

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

Proof:  $p=1$ ,  $p=\infty$  are easy. (i)

Assume that  $1 < p < \infty$ . Also assume that  $u \neq 0$  and  $v \neq 0$ . Then:

$$\begin{aligned} \|u + v\|_p^p &= \sum_{i=1}^n |u_i + v_i|^p \\ &= \sum_{i=1}^n |u_i + v_i|^{p-1} |u_i + v_i| \end{aligned}$$



$$\leq \sum_{i=1}^n |u_i + v_i|^{p-1} (|u_i| + |v_i|) \quad \left( \begin{array}{l} \text{Triangle} \\ \text{inequality} \end{array} \right).$$

Note that By Hölder's inequality, we can write:

$$\sum_{i=1}^n |u_i + v_i|^{p-1} |u_i| \leq \|(\underline{u} + \underline{v})^{p-1}\|_{\frac{p}{p-1}} \|\underline{u}\|_p.$$

$$= \left( \sum_{i=1}^n (u_i + v_i)^p \right)^{\frac{p-1}{p}} \left( \sum_{i=1}^n |u_i|^p \right)^{\frac{1}{p}}.$$

$$= \|\underline{u} + \underline{v}\|_p^{p-1} \|\underline{u}\|_p.$$

Thus, we can continue:

$$\|\underline{u} + \underline{v}\|_p^p \leq \|\underline{u} + \underline{v}\|_p^{p-1} (\|\underline{u}\|_p + \|\underline{v}\|_p).$$

Dividing through by  $\|\underline{u} + \underline{v}\|_p^{p-1}$ , we obtain the inequality.

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## Picture For Norms:

Consider the sets of vectors in  $\mathbb{R}^2$  defined as follows:

$$S'_1 = \{ \underline{v} \in \mathbb{R}^2 : \|\underline{v}\|_1 = 1 \}$$

$$S'_2 = \{ \underline{v} \in \mathbb{R}^2 : \|\underline{v}\|_2 = 1 \}$$

$$S'_\infty = \{ \underline{v} \in \mathbb{R}^2 : \|\underline{v}\|_\infty = 1 \}$$

