

Norms

Monday, January 23, 2017 9:17 AM

Norms add the idea of length and distance and similarity to a vector space.

Definition: A norm $\|\cdot\|$ on a vector space S is a mapping: $\|\cdot\| : S \rightarrow \mathbb{R}$

with the following properties for $\underline{x}, \underline{y} \in S$

1. $\|\underline{x}\| \geq 0$ and $\|\underline{x}\| = 0 \Leftrightarrow \underline{x} = \underline{0}$

2. $\|\underline{x} + \underline{y}\| \leq \|\underline{x}\| + \|\underline{y}\|$ (triangle inequality)

3. $\|a \cdot \underline{x}\| = |a| \cdot \|\underline{x}\|$ for any scalar a
↳ absolute value

The length of $\underline{x} \in S$ is $\|\underline{x}\|$

This distance between \underline{x} and \underline{y} is $\|\underline{x} - \underline{y}\|$

A linear vector space with an associated norm is called a normed vector space.

Examples

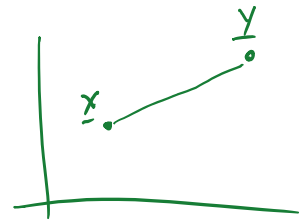
1. $S = \mathbb{R}^N$ - d_2 or Euclidean norm

$$\|\cdot\| = \left(\sum_{i=1}^N |x_i|^2 \right)^{1/2}$$

y

$$\|\underline{x}\|_2 = \left(\sum_{n=1}^N |x_n|^2 \right)^{1/2}$$

corresponds to energy or power (when squared)

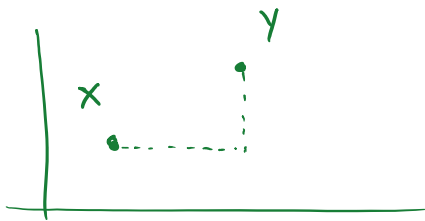


$$\|x-y\|_2 = \sqrt{(x_1-y_1)^2 + (x_2-y_2)^2}$$

2. $S = \mathbb{R}^N$

$$\|\underline{x}\|_1 = \sum_{n=1}^N |x_n|$$

l_1 norm or "taxicab norm" or "Manhattan norm"

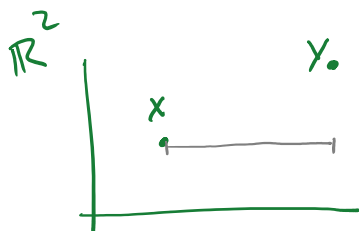


$$\|x-y\|_1 = |x_1-y_1| + |x_2-y_2|$$

3. $S = \mathbb{R}^N$

$$\|\underline{x}\|_\infty = \max_{n=1, \dots, N} |x_n|$$

The l_∞ norm or "Chebyshev norm"



$$\|\underline{x}-\underline{y}\|_\infty = \max(|x_1-y_1|, |x_2-y_2|)$$

4. $S = \mathbb{R}^N$

$$\|\underline{x}\|_p = \left(\sum_{n=1}^N |x_n|^p \right)^{1/p}$$

for some $1 \leq p < \infty$

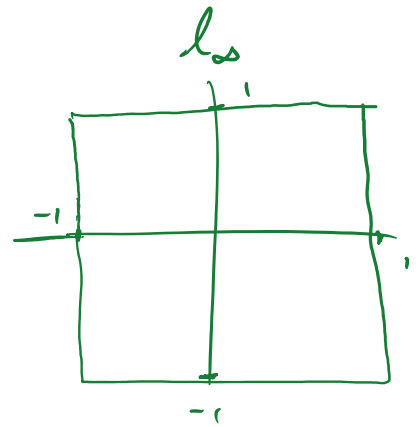
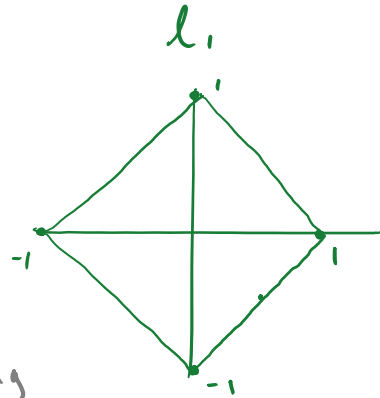
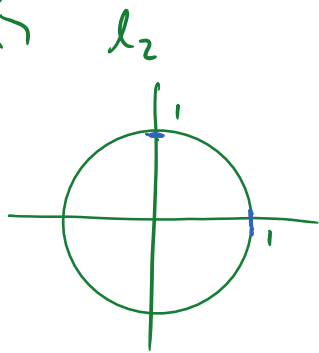
balls in \mathbb{R}^2
 l_1 l_2

This is the l_p -norm

l_1

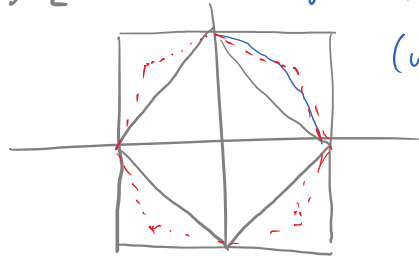
l_∞

unit balls
norm of \mathbb{R}^2



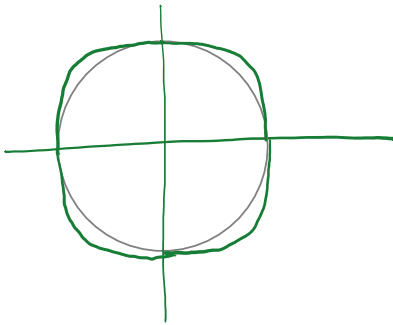
approximating

$l_2 \approx$ average l_1 & l_∞



(weighted average works even better)

l_3 norm



5. $\int =$ continuous-time signals on the real line

$$\|x(t)\|_2 = \left(\int_{-\infty}^{\infty} |x(t)|^2 dt \right)^{1/2}$$

L_2 norm

$\|x(t)\|_2^2$ is the energy in the signal

6. $\int =$ cont.-time functions on $[a, b]$

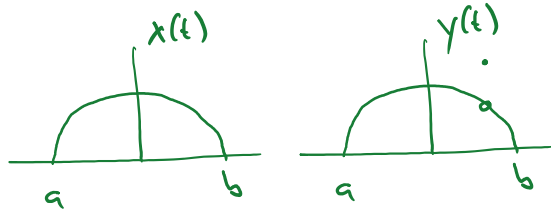
$$\|x(t)\|_p = \left(\int_a^b |x(t)|^p dt \right)^{1/p}$$

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In a normed linear space, we say that

$$\underline{x} = \underline{y} \text{ if } \|\underline{x} - \underline{y}\| = 0$$

for example in $L_2([a, b])$



$\rightarrow y(t)$ differs from $x(t)$ at a single point

then

$$\|\underline{x} - \underline{y}\|_2 = \left(\int_a^b |x(t) - y(t)|^2 dt \right)^{1/2} = 0$$

$x(t) = y(t)$ if they differ at only a finite number of points