

# VECTOR SPACES

A vector space over a field  $k$  (a  $k$ -vector space)

is a set  $V$ , whose elements are called vectors,

with operations  $+$  and  $\cdot$   
addition      scalar multiplication

i.e. given  $v, w \in V$        $v + w \in V$

given  $v \in V, \lambda \in k, \lambda v \in V$

satisfying many rules, e.g.  $v + w = w + v$        $\lambda(v + w) = \lambda v + \lambda w \dots$

E.G.

$$v = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \text{ or } (x_1, \dots, x_n)$$

•  $k$  is a  $k$ -vector space

•  $k^n$  is a  $k$ -vector space

(in particular,  $\mathbb{R}^n$  is an  $\mathbb{R}$ -vector space)

•  $V =$  the set of all functions  $f: \mathbb{R} \rightarrow \mathbb{R}$

•  $V_{cts} = \dots \cup$  continuous functions  $f: \mathbb{R} \rightarrow \mathbb{R}$  is a subspace of  $V$   
(need sum of two continuous functions is continuous)

•  $V_b =$  the set of all bounded  $f: \mathbb{R} \rightarrow \mathbb{R}$

•  $\{ \text{function } f: \text{states} \rightarrow \mathbb{R} \}$

•  $\mathbb{R}$  is a vector space over  $\mathbb{Q}$

## NORM

We want a way of measuring how "large" a vector in  $V$  is

E.G.  $V$  be the space of functions on a  $640 \times 640$  grid

	100	177	
	5		-100

$f(x, y)$  specifies the brightness of pixel  $(x, y)$

so each vector in  $V$  is an image

$\mathbb{R}^{409600}$   
 $\mathbb{R}^{640 \times 640}$

Suppose you have stored images  $V_1, V_2, V_3, \dots, V_n$

W a new image

You believe  $w$  is a match for some  $V_i$ , but which one?

Which  $V_i$  is closest to  $w$ ?

When is  $V_i - w$  small?

What could we mean by the size of a vector  $(x_1, \dots, x_n)$  in  $\mathbb{R}^n$ ?

$$\|(x_1, \dots, x_n)\|_2 = (x_1^2 + \dots + x_n^2)^{1/2} \quad L^2 \text{ norm}$$

$$\|(x_1, \dots, x_n)\|_\infty = \max_i |x_i| \quad L^\infty \text{ norm}$$

$$\|(x_1, \dots, x_n)\|_1 = |x_1| + |x_2| + \dots + |x_n| \quad L^1 \text{ norm}$$

$$\|(x_1, \dots, x_n)\|_p = (|x_1|^p + \dots + |x_n|^p)^{1/p} \quad L^p \text{ norm}$$

Def  $V$  vector space /  $\mathbb{R}$

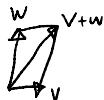
A norm on  $V$  is a function  $\|\cdot\|: V \rightarrow \mathbb{R}$  s.t.

- $\|v\| \geq 0 \quad \forall v \in V$ , and  $\|v\| = 0$  iff  $v = 0$

- $\|\lambda v\| = |\lambda| \|v\|$  for all  $\lambda \in \mathbb{R}, v \in V$

- $\|v+w\| \leq \|v\| + \|w\| \quad \forall v, w \in V$

e.g. for the  $L^2$  norm, says:



What about  $L^\infty$  norm on  $\mathbb{R}^n$

$$v = (x_1, \dots, x_n) \quad w = (y_1, \dots, y_n)$$

$$v+w = (x_1+y_1, \dots, x_n+y_n)$$

To show,

$$\max_i |x_i + y_i| \leq \max_i |x_i| + \max_i |y_i|$$

Let  $j$  in  $[1, \dots, n]$  be the index s.t.

$$|x_j + y_j| = \max_i |x_i + y_i|$$

$$j = \operatorname{argmax}_i |x_i + y_i|$$

$$|v+w|_\infty = \max_i |x_i + y_i| = |x_j + y_j| \leq |x_j| + |y_j|$$

$$\leq \max_i |x_i| + \max_i |y_i|$$

$$= |v|_\infty + |w|_\infty$$

$$\begin{bmatrix} - \\ - \end{bmatrix} \begin{bmatrix} - \\ - \end{bmatrix} \begin{bmatrix} x_i + y_i \end{bmatrix}$$

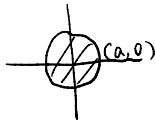
$\mathbb{R}^k$ : the set of functions  $f: \mathbb{R} \rightarrow \mathbb{R}$  bounded between  $-1$  and  $1$  are not a vector space  
 $\sin \in V$  but their sum  $2\sin$  is not in  $V$   
 $\sin \in V$

$V$  vector space  
 $\|\cdot\|$  a norm on  $V$

Def The norm ball  $B_{V, \|\cdot\|}(a)$  is the set  $\{v \in V : \|v\| \leq a\}$

Norm balls in  $\mathbb{R}^2$

$B_{\mathbb{R}^2, \|\cdot\|_2}(a)$



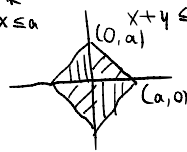
$$(x^2 + y^2)^{1/2} \leq a$$

$$x^2 + y^2 \leq a^2$$

$\mathbb{R}, \|\cdot\|_1, a$



$B_{\mathbb{R}^2, \|\cdot\|_1}(a)$

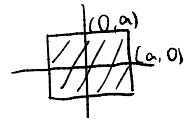


$$|x| + |y| \leq a$$

$\mathbb{R}, \|\cdot\|_\infty$

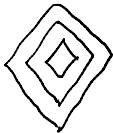


$B_{\mathbb{R}^2, \|\cdot\|_\infty}(a)$



$$\max(|x|, |y|) \leq a$$

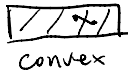
$L_1$  balls for different  $a$



Def We say a subset  $S$  of a vector space  $V$  is convex if, for every  $v, w$  in  $S$ , the line segment  $\overline{vw}$  is contained in  $S$



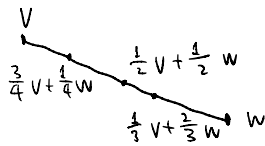
non-convex



convex

Fact: Norm balls are always convex

$$\|\lambda v\| = |\lambda| \|v\|$$



every point on this line segment  
 $\lambda v + (1-\lambda)w \quad \lambda \in [0, 1]$

Pf that norm balls are convex:

Suppose  $v, w \in B_{v, \|\cdot\|}(a)$ . To show:

$$\|\lambda v + (1-\lambda)w\| \leq a \quad (\text{given } \|v\|, \|w\| \leq a)$$

$$\begin{aligned} \|\lambda v + (1-\lambda)w\| &\leq \|\lambda v\| + \|(1-\lambda)w\| \\ &= |\lambda| \|v\| + |1-\lambda| \|w\| \\ &\leq \lambda \|v\| + (1-\lambda) \|w\| \\ &\leq \lambda a + (1-\lambda)a = a \quad \checkmark \end{aligned}$$

Are there good norms on space of  $f: \mathbb{R} \rightarrow \mathbb{R}$   
... continuous functions?

... bounded functions?

Def Let  $V_b$  space of bounded functions  $f: \mathbb{R} \rightarrow \mathbb{R}$

The sup norm on  $V_b$  is defined by

$$\|f\|_{\text{sup}} = \sup_x |f(x)|$$

Why does this exist?

$$E_f = \{|f(x)| \mid x \in \mathbb{R}\}$$

$E_f$  is nonempty  $|f(7)| \in E_f$

$E_f$  is bounded above because  $f$  is bounded

$\Rightarrow$  By LUBP,  $E_f$  has a supremum

To show this is a norm, need to show

$$\|f+g\|_{\text{sup}} \leq \|f\|_{\text{sup}} + \|g\|_{\text{sup}}$$

$$\mathbb{R}^n = \{(x_1, \dots, x_n)\}$$

$$V = \{(f(0), f(1), f(2), f(1.79), f(\pi), \dots)\}$$

What I'd like to say is:

let  $x_0$  be that real # such that

$$|f(x_0)| = \sup_x |f(x)|$$

$$= \text{"arg sup } |f| \text{"}$$

$$f(x) = 1 - \frac{1}{(x^2+1)}$$

$$\|f\|_{\text{sup}} = 1$$



A natural question:

is there some analogue of  $L^2$  norm for functions? Yes

See inner product spaces