

# Convex Sets (following B&V)

Tuesday, January 19, 2021 10:34 PM

Chapter 2 in Boyd & Vandenberghe's book

## §2.1 Affine and convex sets

Def Let  $x, y \in \mathbb{R}^n$  (or any vector space), then

1)  $t x + (1-t)y$ ,  $t \in [0,1]$  is a convex combination (of  $x, y$ )

2)  $t x + (1-t)y$ ,  $t \in \mathbb{R}$  is a linear combination (of  $x, y$ )

3)  $t x + s y$ ,  $t > 0, s > 0$  is a conic combination (of  $x, y$ )

and more generally, for  $K$  points  $\{x_i\}_{i=1}^K$ ,  $\sum_{i=1}^K t_i x_i$  is a ...  
convex combination if  $\sum t_i = 1$ ,  $t_i \geq 0$  a more fitting  
name would be  
"convex cone" combo  
linear combination if  $\sum t_i = 1$  + updated notes  
1/31/21 to fix  
earlier errors  
conic combination if  $t_i \geq 0$ .

Def A set  $C \subseteq \mathbb{R}^n$  is

1) convex if  $\forall x, y \in C$ , it contains all convex combinations of  $x, y$

2) \* affine if  $\forall x, y \in C$ , it contains all linear combinations of  $x, y$

3) a cone if  $\forall x \in C$ , it contains the ray  $\{sx : s \geq 0\}$

4) a convex cone if it's convex and a cone, i.e.,  $\forall x, y \in C$ , contains all conic combinations of  $x, y$   
or equiv.,  $\forall x, y \in C$ ,  $\forall t, s \geq 0$ ,  $tx + sy \in C$

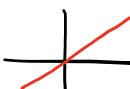
↳ Ex: convex cone   
(more examples later)

non-convex cones

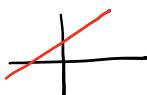


\* An affine set (aka affine subspace) is like a subspace (i.e., a linear subspace) except it's possible shifted (so need not include 0)

Ex: 1D subspace in  $\mathbb{R}^2$  looks like



a 1D affine space in  $\mathbb{R}^2$  might look like



Usually "affine" properties very similar to "linear", and people often say "linear" when it would technically be correct to say "affine"

Ex: dim(affinespace) easily defined by shifting the space (pick  $w_0 \in W$ ,  
 $W$   $V := \{x - w_0 : x \in W\}$ ),  $V$  is now linear,  
define  $\dim(W) = \dim(V)$ .

Just like in real analysis we define things like

$$\text{"closure"} \quad (\bar{A} := \text{set } A \text{ together w/ all limit points} \\ = \text{smallest closed set containing } A \\ = \text{intersection of all closed sets containing } A)$$

we can define:

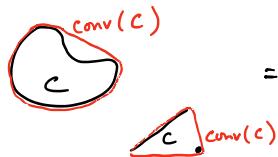
affine hull,  $\text{aff}(C) :=$  smallest affine set containing  $C$

and

$$\text{convex hull}, \text{conv}(C) := \text{smallest convex set containing } C \\ = \text{intersection of all convex sets containing } C \\ = \left\{ t_1x_1 + t_2x_2 + t_3x_3 + \dots + t_kx_k : x_i \in C, \right. \\ \left. t_i \geq 0, \sum_{i=1}^k t_i = 1 \right\}$$

*sometimes/often  
this is the definition  
of convex combination*

= "stretch rubber band around  $C$ ,  
include everything inside"



\* for "any  $k \in \mathbb{N}$ " though  
you can prove  $k=n+1$  is  
sufficient if  $C \subseteq \mathbb{R}^n$

and conic hull is smallest conic and convex set containing  $C$ , etc.

Recall, interior of a set,  $\text{int}(C) := \{x \in C : \exists \varepsilon > 0, B_\varepsilon(x) \subseteq C\}$

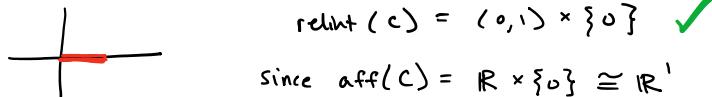
New concept: relative interior of a set,  $\text{relint}(C) := \{x \in C : \exists \varepsilon > 0, B_\varepsilon(x) \cap \text{aff}(C) \subseteq C\}$   
or  $\text{ri}(C)$

makes geometric sense

Ex: let  $C = [0, 1] \subseteq \mathbb{R}^1$ ,  $\text{int}(C) = (0, 1)$ .  
 $\text{relint}(C) = (0, 1)$ .

but *same shape, but embedded in higher dimension*

let  $C = [0, 1] \times \{0\} \subseteq \mathbb{R}^2$ ,  $\text{int}(C) = \emptyset$



$\text{relint}(C) = (0, 1) \times \{0\}$  ✓

since  $\text{aff}(C) = \mathbb{R} \times \{0\} \cong \mathbb{R}^1$

Ex. of common geometric objects

Hyperplane For  $a \in \mathbb{R}^n, b \in \mathbb{R}$ ,  $\{x \in \mathbb{R}^n : a^T x = b\}$ .  $n-1$  dimensional, affine

$n=2$ , these are lines  
 $n=3$ , these are planes

They split  $\mathbb{R}^n$  into 2 sides

Half-space  $\{x \in \mathbb{R}^n : a^T x \leq b\}$  (or another parameterization,  $a^T(x - x_0) \leq 0$   
i.e.,  $b = a^T x_0$ )  
Not affine (so no "dimension")  
but convex

Euclidean ball

open ball  $B_\varepsilon(x) = \{y \in \mathbb{R}^n : \|y - x\| < \varepsilon\}$

Euclidean norm unless otherwise specified

\* oddly,  $B + V$  define  $B_\varepsilon(x)$  to

be the closed ball by default,  
nonstandard in real analysis.  
We won't follow  $B + V$

closed ball  $\overline{B}_\varepsilon(x) = \{y \in \mathbb{R}^n : \|y - x\| \leq \varepsilon\}$

## Ellipsoid

$$E = \{ x : (x - x_0)^T P^{-1} (x - x_0) \leq 1 \}$$

for some matrix  $P > 0$

means  $P$  is positive definite (see supplemental notes)

Includes balls

(choose  $P = \epsilon^2 I$ )

Always convex

## Cones

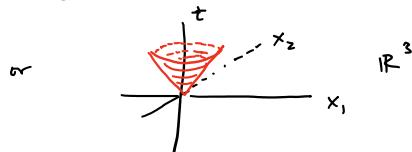
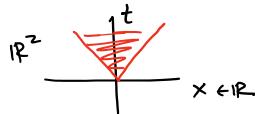
- Positive orthant in  $\mathbb{R}^n$ ,  $\mathbb{R}_+^n = \{ x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \in \mathbb{R}^n : x_i \geq 0 \}$ , is a cone.



$\mathbb{R}_{++}^n = \{ \dots : x_i > 0 \}$  isn't a cone  
(doesn't include 0, but 0 is always a convex combination)

- Lorentz cone / 2nd order cone / "ice cream cone"

$$C = \{ (x, t) \in \mathbb{R}^{n+1}, \begin{array}{l} x \in \mathbb{R}^n \\ t \in \mathbb{R} \end{array} : \|x\|_2 \leq t \}$$



PSD

- Positive Semidefinite matrices. Most important non-polyhedral cone

Recall  $\mathbb{R}^{m \times n} = \{ m \times n \text{ matrices} \}$  is a  $m \times n$  dimensional linear space  
aka vector space

and let  $S^n = \{ X \in \mathbb{R}^{n \times n} : X = X^T \}$  be square, symmetric matrices  
(or  $X = X^*$ , if working w/  $\mathbb{C}^{n \times n}$ )

⚠ (often  $S^{n-1}$  means  
the sphere in  $\mathbb{R}^n$ ,  
but we'll usually  
use it to mean  
symm. matrices)

$S^n$  is a linear space

Define PSD cone to be  $S_+^n = \{ X \in S^n : X \geq 0 \}$

(Pos. Def. set  $S_{++}^n = \{ X \in S^n : X > 0 \}$   
isn't a cone and isn't closed)

means  $X$  is psd

## Polyhedron

⚠ Very common in math so there are conflicting definitions

Def: polyhedron  $P \subseteq \mathbb{R}^n$  is a set of the intersection of a finite number  
of half-spaces and half-planes

Equiv def'n:

A set is polyhedral  
if it can be written  
as the convex hull  
of a finite #  
of points

$$P = \{ x \in \mathbb{R}^n : a_j^T x \leq b_j, j=1, \dots, m \}$$

$$c_j^T x = d_j, j=1, \dots, p \}$$



⚠  $\infty$  # half-spaces/planes isn't necessarily a polyhedron

$$\text{Ex: } \{ x \in \mathbb{R}^n : a_j^T x \leq 1, a_j \in \text{unit circle} \}$$

defines the closed ball  $\overline{B_r(0)}$ , not polyhedral

Polyhedra are always convex

Some authors call a bounded polyhedron a polytope

(others define a polytope as we did a polyhedra, and define a polyhedra to mean a bounded polytope! completely switched!)

In other fields / wikipedia,

polygon	means in $\mathbb{R}^2$	We won't use these definitions.
polyhedron	means in $\mathbb{R}^3$	
polytope	means higher dimension.	

I'll probably use all 3 interchangeably... sorry!

To prove a set is polyhedral...

- 1) show it's the finite intersection of polyhedra
- ~ 2) show it can be written as  $\text{conv}(V)$  for a finite set of vertices  $V$
- 3) lift to a higher dimensional space, show that set is polyhedral, and the original set is the projection of that polyhedron.