



Lines and Line segments

* $y = x_2 + \theta(x_1 - x_2)$

Affine sets

* A set $C \subseteq \mathbb{R}^n$ is affine, if line through any two distinct points in C lies in C .

Affine sets * $y = \theta x_1 + (1-\theta)x_2$ $\theta \in \mathbb{R}$ and $x_1, x_2 \in C$

Affine combination * $\theta_1 x_1 + \theta_2 x_2 + \dots + \theta_k x_k \in C$; $\theta_1 + \theta_2 + \dots + \theta_k = 1$

Subspace * $V = C - x_0 = \{x - x_0 | x \in C\}$; $x_0 \in C$
dim. of $C = \text{dim. of } V$

$\{x | Ax = b\} \rightarrow \text{sol}^n$ set of sys. of LE is affine, subspace = nullspace of A .

Affine hull * The set of all affine combination of points in C .

'aff C '. If S is any affine set with $C \subseteq S$, aff $C \subseteq S$.

Affine dim \rightarrow dimension of affine hull

CONVEX SETS

* A set C is convex if the line segment b/w any two points in C lies in C , i.e. if for any $x_1, x_2 \in C$ and any θ with $0 \leq \theta \leq 1$, we have

$\theta x_1 + (1-\theta)x_2 \in C$

CONVEX COMBINATION

* $\theta_1 x_1 + \theta_2 x_2 + \dots + \theta_k x_k$ $\theta_1 + \theta_2 + \dots + \theta_k = 1$; $\theta_i \geq 0$

CONVEX HULL

'conv C ' - the smallest convex set that contains C .
the set of all convex combination of points in C .

CONES

* A set C is called cone / non-negative homogeneous, if for every $x \in C$ and $\theta \geq 0$ we have $\theta x \in C$.

Convex cone - $\theta_1 x_1 + \theta_2 x_2 \in C$ $x_1, x_2 \in C$ & $\theta_1, \theta_2 \geq 0$.

Conic combination * $\theta_1 x_1 + \theta_2 x_2 + \dots + \theta_k x_k$ $\theta_1, \theta_2, \dots, \theta_k \geq 0$.

conic hull the set of all conic combinations of points in C .

i.e. smallest convex cone that contains C .

open HS $\{x | a^T x < b\}$

* all vectors that have non-negative inner product with K .

DUAL CONES

K^* always convex and $K^* = \{y | x^T y \geq 0 \forall x \in K\}$
is a cone though K is not

* dual cone of subspace is its orthogonal complement

SEPARATING & SUPPORTING HP

* Separating hyperplane theorem:
 $C \cap D = \emptyset$
 $a^T x \leq b$ $a^T x \geq b$

ordering in \mathbb{R}^n space

GENERALIZED INEQUALITIES

as cone is directional we use it for comparison

OPERATIONS THAT PRESERVE CONVEXITY

CONVEX SETS

NORM CONE

$C = \{(x, t) | \|x\| \leq t\} \subseteq \mathbb{R}^{n+1}$

* when $\|\cdot\|_2 \Rightarrow$ 2nd order cone \Rightarrow quadratic cone \Rightarrow Lorentz cone \Rightarrow ice-cream cone

HYPERPLANES AND HALFSPACES

$\{x | a^T x = b\}$, $a \in \mathbb{R}^n$, $a \neq 0$ and $b \in \mathbb{R}$

normal vector
offset from origin
hyperplane contains offset x_0 , plus all vectors orthogonal to the normal vector a .

$\{x | a^T x \leq b\}$

IHS
IHP * closed HS
IHS

x_0 plus any vector that makes an obtuse angle with normal vector a .

* if C is a convex set, then every boundary points of it contains a HP that clearly separates the convex set into 2 HS.

* Such HP called supporting HP.

$K \subseteq \mathbb{R}^n$ Proper cone
 $x \leq_K y \Leftrightarrow y - x \in K$

Intersection

Polyhedron - $HS + IHP$

PSD \Rightarrow HS $\{x \in \mathbb{R}^n | z^T x z \geq 0\}$

Affine function

$f(x) = Ax + b$; if $S \subseteq \mathbb{R}^n$ is convex then image of f under S ($f(S)$) is also convex.

* scaling, translation.

Perspective func.

$P: \mathbb{R}^{n+1} \rightarrow \mathbb{R}^n$

last component is one.

$P(z, t) = z/t$

Linear fractional func.

$f = P \circ a$

k -dimensional simplex in \mathbb{R}^n

$C = \text{conv}\{v_0, v_1, \dots, v_k\}$

The affine dimension of this $(k+1)$ affinely indep. points is k .

* By introducing simplex, we were able to increase a dimension to convex set.

non-negative orthant is a polyhedral cone

* bounded n - polytope

polyhedron

polyhedron

polyhedron