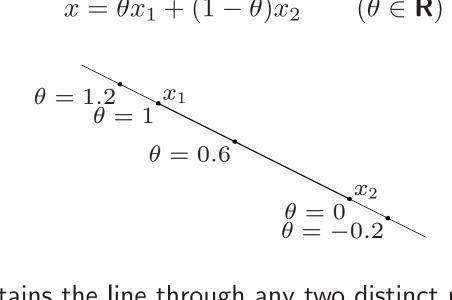
2. Convex sets

- affine and convex sets
- some important examples
- operations that preserve convexity
- generalized inequalities
- separating and supporting hyperplanes
- dual cones and generalized inequalities

Affine set

line through x_1 , x_2 : all points

$$x = \theta x_1 + (1 - \theta) x_2 \qquad (\theta \in \mathbf{R})$$



affine set: contains the line through any two distinct points in the set

example: solution set of linear equations $\{x \mid Ax = b\}$

(conversely, every affine set can be expressed as solution set of system of linear equations)

Convex set

line segment between x_1 and x_2 : all points

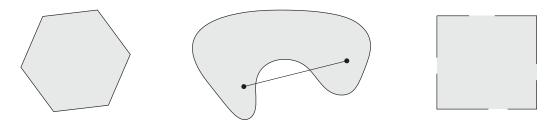
$$x = \theta x_1 + (1 - \theta)x_2$$

with $0 \le \theta \le 1$

convex set: contains line segment between any two points in the set

$$x_1, x_2 \in C, \quad 0 \le \theta \le 1 \quad \Longrightarrow \quad \theta x_1 + (1 - \theta)x_2 \in C$$

examples (one convex, two nonconvex sets)



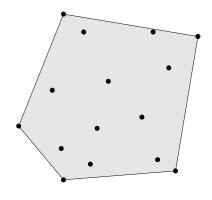
Convex combination and convex hull

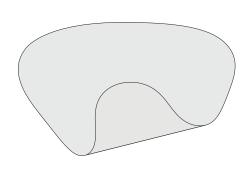
convex combination of x_1, \ldots, x_k : any point x of the form

$$x = \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_k x_k$$

with
$$\theta_1 + \cdots + \theta_k = 1$$
, $\theta_i \ge 0$

convex hull conv S: set of all convex combinations of points in S



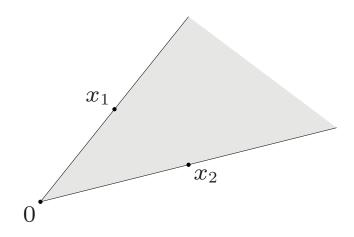


Convex cone

conic (nonnegative) combination of x_1 and x_2 : any point of the form

$$x = \theta_1 x_1 + \theta_2 x_2$$

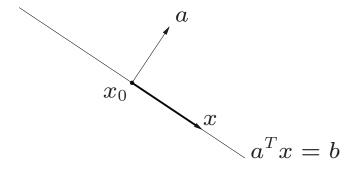
with $\theta_1 \ge 0$, $\theta_2 \ge 0$



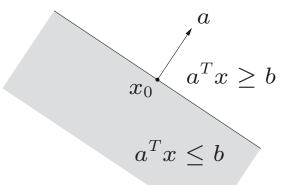
convex cone: set that contains all conic combinations of points in the set

Hyperplanes and halfspaces

hyperplane: set of the form $\{x \mid a^T x = b\}$ $(a \neq 0)$



halfspace: set of the form $\{x \mid a^T x \leq b\}$ $(a \neq 0)$



- ullet a is the normal vector
- hyperplanes are affine and convex; halfspaces are convex

Euclidean balls and ellipsoids

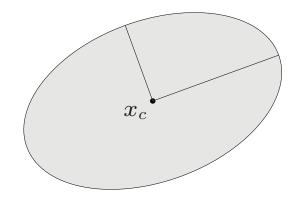
(Euclidean) ball with center x_c and radius r:

$$B(x_c, r) = \{x \mid ||x - x_c||_2 \le r\} = \{x_c + ru \mid ||u||_2 \le 1\}$$

ellipsoid: set of the form

$$\{x \mid (x - x_c)^T P^{-1} (x - x_c) \le 1\}$$

with $P \in \mathbf{S}_{++}^n$ (i.e., P symmetric positive definite)



other representation: $\{x_c + Au \mid ||u||_2 \le 1\}$ with A square and nonsingular

Norm balls and norm cones

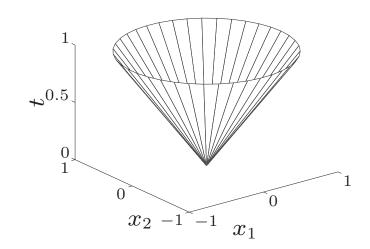
norm: a function $\|\cdot\|$ that satisfies

- $||x|| \ge 0$; ||x|| = 0 if and only if x = 0
- ||tx|| = |t| ||x|| for $t \in \mathbf{R}$
- $||x + y|| \le ||x|| + ||y||$

notation: $\|\cdot\|$ is general (unspecified) norm; $\|\cdot\|_{\text{symb}}$ is particular norm **norm ball** with center x_c and radius r: $\{x\mid \|x-x_c\|\leq r\}$

norm cone: $\{(x,t) \mid ||x|| \le t\}$

Euclidean norm cone is called secondorder cone



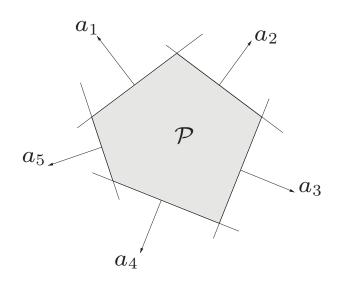
norm balls and cones are convex

Polyhedra

solution set of finitely many linear inequalities and equalities

$$Ax \leq b, \qquad Cx = d$$

 $(A \in \mathbf{R}^{m \times n}, C \in \mathbf{R}^{p \times n}, \leq \text{is componentwise inequality})$



polyhedron is intersection of finite number of halfspaces and hyperplanes

Positive semidefinite cone

notation:

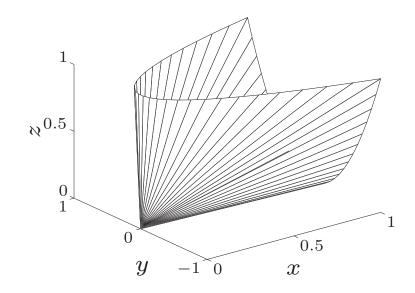
- S^n is set of symmetric $n \times n$ matrices
- $\mathbf{S}_{+}^{n} = \{X \in \mathbf{S}^{n} \mid X \succeq 0\}$: positive semidefinite $n \times n$ matrices

$$X \in \mathbf{S}^n_+ \iff z^T X z \ge 0 \text{ for all } z$$

 \mathbf{S}^n_+ is a convex cone

• $\mathbf{S}_{++}^n = \{X \in \mathbf{S}^n \mid X \succ 0\}$: positive definite $n \times n$ matrices

example: $\begin{bmatrix} x & y \\ y & z \end{bmatrix} \in \mathbf{S}_{+}^{2}$



Operations that preserve convexity

practical methods for establishing convexity of a set C

1. apply definition

$$x_1, x_2 \in C, \quad 0 \le \theta \le 1 \quad \Longrightarrow \quad \theta x_1 + (1 - \theta)x_2 \in C$$

- 2. show that C is obtained from simple convex sets (hyperplanes, halfspaces, norm balls, . . .) by operations that preserve convexity
 - intersection
 - affine functions
 - perspective function
 - linear-fractional functions