

Section 2.5 Relations to convex geometry

Background on convex sets

Def'n A set C is said to be convex if for every $x_1, x_2 \in C$ and every λ with $0 < \lambda < 1$, $\lambda x_1 + (1-\lambda)x_2 \in C$.

line or
line segment
convex



• intersection of convex sets is convex (follows quickly from def'n)
but not the union

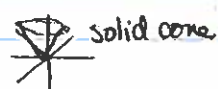
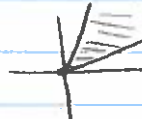


Def'n The convex hull of a set S , denoted $co(S)$, is the intersection of all convex sets containing S .
Can also think of as the set of all convex combinations of points in S : $co(S) = \{x = \sum_{i=1}^n \lambda_i x_i : \lambda_i \geq 0, \sum \lambda_i = 1, x_i \in S\}$

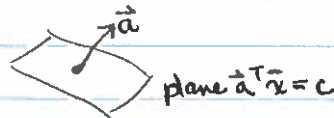


Def'n A set C is a cone if $x \in C$ implies $\alpha x \in C$ for all $\alpha > 0$.
A cone that is also convex is called a convex cone.

Examples: line through origin



Def'n $\overset{\text{open}}{H}_+ = \{ \vec{x} : \vec{a}^T \vec{x} > c \}$ closed $H_+ \geq$
 $\overset{\text{open}}{H}_- = \{ \vec{x} : \vec{a}^T \vec{x} < c \}$ $H_- \leq$



Example: $\vec{a} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, c = 4$ $H_+ = \{ \begin{bmatrix} 1 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} x \\ y \end{bmatrix} \geq 4 \} = \{ x + 2y \geq 4 \}$



Def'n A set that is the intersection of a finite # of closed half spaces is called a convex polytope.



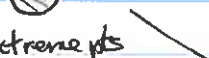
(set of solns to system of inequalities $\vec{a}_1^T \vec{x} \leq b_1, \dots, \vec{a}_m^T \vec{x} \leq b_m$)

Worksheet on convex sets

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Def'n An extreme point x of a convex set C is such that there are no two distinct pts x_1 and x_2 in C s.t.
 $x = \alpha x_1 + (1-\alpha)x_2$ for some α with $0 < \alpha < 1$

Examples: corners of polytopes
 boundary of disk
 a line has no extreme pts

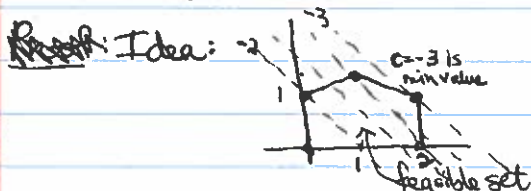


Thm: Let A be an $m \times n$ matrix of rank m and $b \in \mathbb{R}^m$.

Let K be the convex polytope consisting of all $\bar{x} \in \mathbb{R}^n$ satisfying $A\bar{x} = \bar{b}$, $\bar{x} \geq 0$.

Then a vector \bar{x} is an extreme pt of K iff \bar{x} is a basic feasible soln.

* Reduces solving LPs to looking at corner pts in typical problems.



level curves of objective fn $-x-y$
 are lines $-x-y=c$, $c = \text{value of obj fn at } (x,y)$
 $y = -x - c$

Proof: " \Leftarrow " Suppose $\bar{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ is a basic feasible soln to $A\bar{x} = \bar{b}$, $\bar{x} \geq 0$, with k columns of A lin. ind.

Then $A\bar{x} = x_1 \bar{a}_1 + \dots + x_k \bar{a}_k = \bar{b}$. We need to show \bar{x} is an extreme pt.

Suppose it is not: $\bar{x} = \alpha \bar{y} + (1-\alpha)\bar{z}$, $0 < \alpha < 1$, $\bar{y} \neq \bar{z}$ in K . All components of \bar{x} , \bar{y} , and \bar{z} are nonneg, as is α , the last $n-k$ components of \bar{y} & \bar{z}

$$\begin{aligned} (x_1 - y_1)\bar{a}_1 + \dots + (x_k - y_k)\bar{a}_k &= \bar{0} \\ (y_1 - z_1)\bar{a}_1 + \dots + (y_k - z_k)\bar{a}_k &= \bar{0} \\ \Rightarrow x_k - y_k = 0, y_k - z_k = 0 \end{aligned}$$

are zero (as in \bar{x}). So $y_1 \bar{a}_1 + \dots + y_k \bar{a}_k = \bar{b} = z_1 \bar{a}_1 + \dots + z_k \bar{a}_k$.

Because $\bar{a}_1, \dots, \bar{a}_k$ are lin. ind., $\bar{x} = \bar{y} = \bar{z}$, contradicting our assumption.

Hence \bar{x} is an extreme pt.

" \Rightarrow " Suppose \bar{x} is an extreme pt of K . WLOG assume the nonzero components of \bar{x} are x_1, \dots, x_k , so $x_1 \bar{a}_1 + \dots + x_k \bar{a}_k = \bar{b}$. We want to show \bar{x} is a basic feasible soln, which means we need to show $\{\bar{a}_1, \dots, \bar{a}_k\}$ lin. ind.

Suppose not: $\exists y_1, \dots, y_k$ not all zero s.t. $y_1 \bar{a}_1 + \dots + y_k \bar{a}_k = \bar{0}$.

Let $\bar{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_k \\ 0 \\ \vdots \\ 0 \end{bmatrix}$. Because $x_1, \dots, x_k > 0$, we can find $\epsilon > 0$ s.t. $\bar{x} \pm \epsilon \bar{y} \geq 0$.

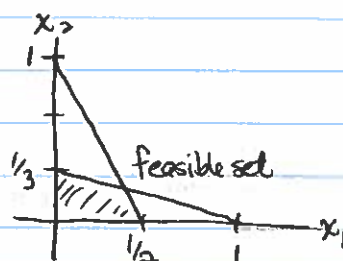
Then $\bar{x} = \frac{1}{2}(\bar{x} + \epsilon \bar{y}) + \frac{1}{2}(\bar{x} - \epsilon \bar{y})$. $A(\bar{x} \pm \epsilon \bar{y}) = \bar{b} + \epsilon \bar{0} = \bar{b}$, so

$\bar{x} \pm \epsilon \bar{y} \in K$, making \bar{x} not an extreme pt, a contradiction.

Hence \bar{x} is a basic feasible soln. \square

Exploratory example:

$$\min_{\vec{x} \in \mathbb{R}^2} \vec{c}^T \vec{x} \text{ subject to } \begin{aligned} 2x_1 + x_2 &\leq 1 \\ x_1 + 3x_2 &\leq 1 \\ x_1, x_2 &\geq 0 \end{aligned}$$



Standard form: $2x_1 + x_2 + x_3 = 1$

$$x_1 + 3x_2 + x_4 = 1$$

$$x_1, x_2, x_3, x_4 \geq 0$$

$$A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 1 & 3 & 0 & 1 \end{bmatrix}^* \text{ rank } A = 2$$

$$\vec{b} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Basic feasible sol's are those with at least 2 zeros

- rearrange columns of A to have different pairs of lin ind columns as first two columns, row reduce $[A|b]$ to solve, set free variables to zero (and put in original order)

$\binom{4}{2} = 6$ ways

R script

$$\begin{bmatrix} 2/5 \\ 1/5 \\ 0 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 1 \\ 0 \\ -1 \\ 0 \end{bmatrix} \quad x_3 < 0$$

$$\begin{bmatrix} 1/2 \\ 0 \\ 0 \\ 1/2 \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ 1/3 \\ 2/3 \\ 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ -2 \end{bmatrix} \quad x_4 < 0$$

$$\begin{bmatrix} 0 \\ 0 \\ 0 \\ 1 \end{bmatrix}$$

Correspond to 4 extreme pts

$$\vec{c} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}: \min x_1 + x_2 \quad \text{sol'n } \vec{x} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}, \text{ min value } 0$$

$$\vec{c} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}: \min -x_1 - x_2 \quad \vec{x} = \begin{bmatrix} 2/5 \\ 1/5 \end{bmatrix}, \text{ min value } -3/5$$

$$\vec{c} = \begin{bmatrix} -1 \\ 0 \end{bmatrix}: \min -x_1 \quad \vec{x} = \begin{bmatrix} 1/2 \\ 0 \end{bmatrix}, \text{ min value } -1/2$$

$$\vec{c} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}: \min -x_2 \quad \vec{x} = \begin{bmatrix} 0 \\ 1/3 \end{bmatrix}, \text{ min value } -1/3$$

$$\vec{c} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}: \min x_1 \quad \vec{x} = \begin{bmatrix} 0 \\ x_2 \end{bmatrix}, 0 \leq x_2 \leq 1/3 \quad \text{entire line segment with min value } 0$$

If flip inequalities to \geq , have unbounded region for which there may be no sol'n, e.g., $\vec{c} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$. Or the feasible set could be empty, so no sol'n possible

How to discern these different cases?

Section 2.6 Farkas' Lemma

- way to check whether a feasible sol'n exists for an LP

Thm (Farkas' Lemma)

Let A be an $m \times n$ matrix and $\vec{b} \in \mathbb{R}^m$.

Then $A\vec{x} = \vec{b}, \vec{x} \geq 0$ has a feasible sol'n $\Leftrightarrow -\vec{y}^T A \geq \vec{0}, \vec{y}^T \vec{b} = 1$ has no feasible sol'n \vec{y} .

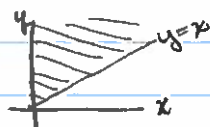
(so finding such a \vec{y} means the original system is infeasible)
for the "alternative" system

Lemma: Let C be the cone generated by the columns of A :

$C = \{A\vec{x} : \vec{x} \geq \vec{0}\}$. Then C is a closed and convex set.

Examples: $A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}$ $A\vec{x} = \begin{bmatrix} x_1 \\ x_1 + x_2 \end{bmatrix} = \begin{bmatrix} x \\ y \end{bmatrix}$

$$y = x_1 + x_2 \geq x$$



$$A = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$$

$$A\vec{x} = \begin{bmatrix} x_1 + x_2 \\ x_1 + x_2 \end{bmatrix} \text{ so } x = y$$



$$A = \begin{bmatrix} 2 & 0 \\ 1 & -1 \end{bmatrix}$$

$$A\vec{x} = \begin{bmatrix} 2x_1 \\ x_1 - x_2 \end{bmatrix} \text{ so } y \leq x/2$$



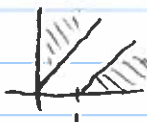
Proof of thm: read pages 33-34.

Geometric interpretation: if \vec{b} is not in this cone C , then

there must be a hyperplane separating \vec{b} and the cone C ,

where sol'n \vec{y} to the alternative system is ~~the normal~~ ^{the normal} vector to the hyperplane.

Example: $x_1 - x_2 \leq 0$
 $-x_1 + x_2 \leq -1$
 $x_1, x_2 \geq 0$



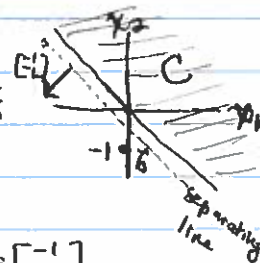
no intersection,
empty feasible set
(constraints are infeasible)

$$A = \begin{bmatrix} -1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

$$\vec{b} = \begin{bmatrix} 0 \\ -1 \end{bmatrix}$$

Standard form: $x_1 - x_2 + x_3 = 0$
 $-x_1 + x_2 + x_4 = -1, x_1, x_2, x_3, x_4 \geq 0$

Cone $C = \{c_1 \begin{bmatrix} -1 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + c_3 \begin{bmatrix} 0 \\ 1 \end{bmatrix} + c_4 \begin{bmatrix} 0 \\ 1 \end{bmatrix} : c_1, c_2, c_3, c_4 \geq 0\}$
any multiple of $\begin{bmatrix} -1 \\ 0 \end{bmatrix}$ can increase x_1 and/or x_2



Alt. system $[y_1, y_2] \begin{bmatrix} -1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} -y_1 + y_2 \\ y_1 - y_2 \end{bmatrix} \geq \begin{bmatrix} 0 \\ 0 \end{bmatrix} \Rightarrow y_1 = y_2$
and $[y_1, y_2] \begin{bmatrix} 0 \\ -1 \end{bmatrix} = 1$
 $-y_2 = 1 \Rightarrow y_2 = -1$
so $\vec{y} = \begin{bmatrix} -1 \\ -1 \end{bmatrix}$