

CS615 2019

Lecture 10: Linear rational arithmetic (basics)

Instructor: Ashutosh Gupta

IITB, India

Compile date: 2019-10-07

Reasoning over linear arithmetic

Example 10.1

Consider the following proof step

$$\frac{2x - y \leq 1 \quad 4y - 2x \leq 6}{x + y \leq 5}$$

Is the above proof step complete?

Basic concepts

One needs to know the following

- ▶ Linearly independent
- ▶ Rank of a set of vectors
- ▶ Vector vs. Row vector
- ▶ Hyperplane
- ▶ Affine hull

Fundamental theorem of linear inequality

Theorem 10.1

Let a_1, \dots, a_m and b be n -dimensional vectors.

Then, one of the following is true.

1. $b := \lambda_1 a_{i_1} + \dots + \lambda_k a_{i_k}$ for $\lambda_j \geq 0$ and a_{i_1}, \dots, a_{i_k} are linearly independent
2. There exists a hyperplane $\{x | cx = 0\}$ containing $t - 1$ linearly independent vectors from a_1, \dots, a_m such that

$$ca_1 \geq 0, \dots, ca_m \geq 0 \text{ and } cb < 0,$$

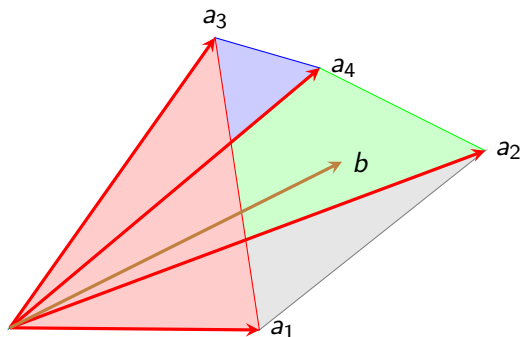
where $t := \text{rank}\{a_1, \dots, a_m, b\}$.

Observation:

- ▶ c is a row vector
- ▶ Wlog, we assume $t = n$. (why?)
- ▶ Both possibilities cannot be true at the same time. (why?)

Geometrically, theorem case 1

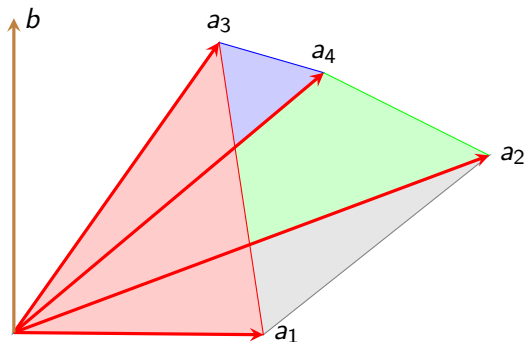
In the first case, b is in the cone of a_1, \dots, a_m .



Geometrically theorem case 2

In the second case, b is outside of the cone of a_1, \dots, a_m .

Furthermore, a_1, \dots, a_m are in one side of $\{x | cx = 0\}$ and b is on the other.



Exercise 10.1

Give a c ?

Proof: fundamental theorem of linear inequality

Proof.

Consider the following iterative algorithm to decide case 1 or 2.

Initially choose n independent vectors $D := \{a_{i_1}, \dots, a_{i_n}\}$ from a_1, \dots, a_m .

1. Let $b = \lambda_{i_1} a_{i_1} + \dots + \lambda_{i_n} a_{i_n}$.
2. If $\lambda_{i_1}, \dots, \lambda_{i_n} \geq 0$, case 1 and exit.
3. Otherwise, choose **smallest** i_h such that $\lambda_{i_h} < 0$.
4. Choose c such that $ca = 0$ for each $a \in D \setminus \{a_{i_h}\}$ and $ca_{i_h} = 1$.
5. If $ca_1, \dots, ca_m \geq 0$, case 2 and exit. (why?)
6. Otherwise, choose smallest s such that $ca_s < 0$.
7. $D := D \setminus \{a_{i_h}\} \cup \{a_s\}$. goto 1.

Clearly, $cb < 0$. (why?)

...

Exercise 10.2

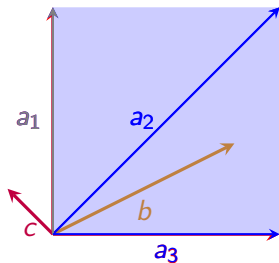
- a. Why λ s exists in step 1?
- b. Why c exists in step 4?
- c. Why D remains linearly independent over time?
- d. Why not simply enumerate all linearly independent subsets from a_1, \dots, a_m ?

Example: iterations for D

Example 10.2

Let us have a set of vectors $\{a_1, a_2, a_3\}$ in 2-dimensional vector space and also vector b . We are looking for a subset D that contains b in its cone.

1. Initial guess, $D = \{a_1, a_2\}$.
2. If we write $b = \lambda_1 a_1 + \lambda_2 a_2$, then $\lambda_1 < 0$.
3. Clearly b is not in the cone of D .
4. We get c such that $ca_2 = 0$ and $ca_1 > 0$.
5. Since $cb = c(\lambda_1 a_1 + \lambda_2 a_2) = \lambda_1 ca_1$, $cb < 0$.
6. We find a_3 such that $ca_3 < 0$
(Intuition: a_3 is likely to be closer to b)
7. Now $D := D \setminus \{a_1\} \cup \{a_3\} = \{a_2, a_3\}$
8. b is in the cone of D . Terminate.



Proof: fundamental theorem of linear inequality II

Proof.

We are yet to prove termination of the algorithm.

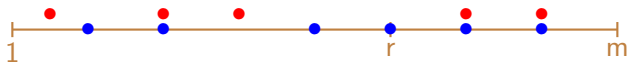
Let D^k be the set D at iteration k .

claim: D^k will not repeat in any future iterations. (Therefore, termination.)

Contrapositive: For some $\ell > k$, $D^\ell = D^k$.

Let r be the highest index such that a_r left D at p th iteration and came back at q th iteration for $k \leq p < q \leq \ell$

Therefore, $D^p \cap \{a_{r+1}, \dots, a_m\} = D^q \cap \{a_{r+1}, \dots, a_m\}$...



Blue dots are indexes for D^p . Red dots are indexes for D^q .

Proof: fundamental theorem of linear inequality III

Proof.

$$D^p := \{a_{i_1^p}, \dots, a_{i_n^p}\}$$

$$\text{Let } b = \lambda_{i_1^p} a_{i_1^p} + \dots + \lambda_{i_n^p} a_{i_n^p}.$$

Since r left D^p ,

if $i_j^p < r$, $\lambda_{i_j^p} \geq 0$ and

if $i_j^p = r$, $\lambda_r < 0$.

At q th iteration, we have $c^q b < 0$.

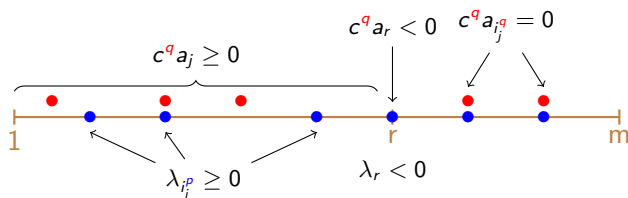
Since r entered in D^q ,

for each $j < r$, $c^q a_j \geq 0$,

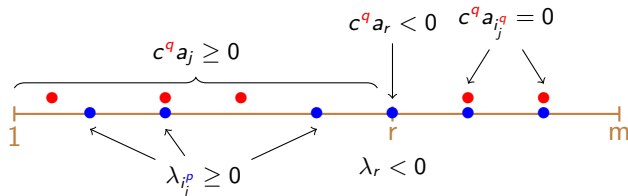
for each $j = r$, $c^q a_r < 0$, and

for each $i_j^q > r$, $c^q a_{i_j^q} = 0$.

...



Proof: fundamental theorem of linear inequality IV



Proof.

Consider

$$0 > c^q b = c^q (\lambda_{i_1^p} a_{i_1^p} + \cdots + \lambda_{i_n^p} a_{i_n^p})$$

Let us show for each j , $\lambda_{i_j^p} (c^q a_{i_j^p})$ is nonnegative.

Three cases

- ▶ $i_j^p < r$: $\lambda_{i_j^p} \geq 0$ and $c^q a_{i_j^p} \geq 0$
- ▶ $i_j^p = r$: $\lambda_r < 0$ and $c^q a_r < 0$
- ▶ $i_j^p > r$: $c^q a_{i_j^p} = 0_{(\text{why?})}$

Therefore, $c^q b \geq 0$. **Contradiction.**



Cone, Polyhedra

Definition 10.1

A set C of vectors is a **cone** if $x, y \in C$ then $\lambda_1 x + \lambda_2 y \in C$ for each $\lambda_1, \lambda_2 \geq 0$.

Definition 10.2

A cone C is a **polyhedral** if $C = \{x | Ax \leq 0\}$ for some matrix A .

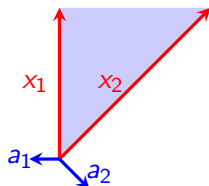
Definition 10.3

A cone C is **finitely generated** by vectors x_1, \dots, x_m is the set

$$\text{cone}\{x_1, \dots, x_m\} := \{\lambda_1 x_1 + \dots + \lambda_m x_m | \lambda_1, \dots, \lambda_m \geq 0\}$$

Example 10.3

$$C = \{x | \begin{bmatrix} a_1 \\ a_2 \end{bmatrix} x \leq 0\} = \{\lambda_1 x_1 + \lambda_2 x_2 | \lambda_1, \lambda_2 \geq 0\}$$



Polyhedra == finitely generated cone

Theorem 10.2

A convex cone is polyhedral iff it is finitely generated.

Proof.

Intuitively, obvious.

We are skipping the proof here.



Polyhedron, affine half space, polytope

Definition 10.4

A set of vectors P is called *polyhedron* if

$$P = \{x | Ax \leq b\}$$

for some matrix A and vector b .

Definition 10.5

A set of vectors H is called *affine half-space* if

$$H = \{x | wx \leq \delta\}$$

for some nonzero row vector w and number δ .

Polytope

Definition 10.6

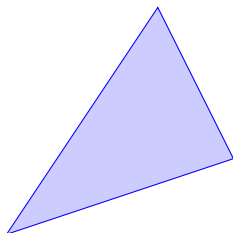
A set of vectors Q is called *polytope* if

$$Q = \text{hull}(\{x_1, \dots, x_m\}) = \{\lambda_1 x_1 + \dots + \lambda_m x_m \mid \lambda_1 + \dots + \lambda_m = 1 \wedge \lambda_1, \dots, \lambda_m \geq 0\}$$

for some nonzero vectors x_1, \dots, x_m .

Example 10.4

The following is $\text{hull}(\{(2, 3), (0, 0), (3, 1)\})$



polyhedron = polytope + polyhedral

Theorem 10.3 (Decomposition theorem)

Let $P = \{x | Ax \leq b\}$ be a polyhedron iff $P = Q + C$ for some polytope Q and polyhedral C .

Proof.

Let us consider the forward direction.

Let us construct the following cone in one higher dimension.

$$P' = \left\{ \begin{bmatrix} x \\ \lambda \end{bmatrix} \mid Ax - \lambda b \leq 0 \wedge \lambda \geq 0 \right\}$$

Clearly, the following holds

$$x \in P \quad \text{iff} \quad \begin{bmatrix} x \\ 1 \end{bmatrix} \in P'$$

Exercise 10.3

Prove the reverse direction

...

polyhedron = polytope + polyhedral (contd.)

Proof(contd.)

Let the following $q + c$ vectors generate P' . (why exists?)

$$\underbrace{\begin{bmatrix} x_1 \\ 1 \end{bmatrix} \cdots \begin{bmatrix} x_q \\ 1 \end{bmatrix}}_q, \underbrace{\begin{bmatrix} y_1 \\ 0 \end{bmatrix} \cdots \begin{bmatrix} y_c \\ 0 \end{bmatrix}}_c$$

Let $Q = \text{hull}(\{x_1, \dots, x_q\})$ and $C = \text{cone}(\{y_1, \dots, y_c\})$

claim: $P = Q + C$

Let $x \in P$

\Leftrightarrow By definition of P' , for some $\mu_1, \dots, \mu_q, \lambda_1, \dots, \lambda_c \geq 0$ the following holds.

$$\begin{bmatrix} x \\ 1 \end{bmatrix} = \mu_1 \begin{bmatrix} x_1 \\ 1 \end{bmatrix} + \cdots + \mu_q \begin{bmatrix} x_q \\ 1 \end{bmatrix} + \lambda_1 \begin{bmatrix} y_1 \\ 0 \end{bmatrix} + \cdots + \lambda_c \begin{bmatrix} y_c \\ 0 \end{bmatrix}.$$

$\Leftrightarrow \mu_1 x_1 + \dots + \mu_q x_q \in Q$ and $\lambda_1 y_1 + \dots + \lambda_c y_c \in C$ (why?)

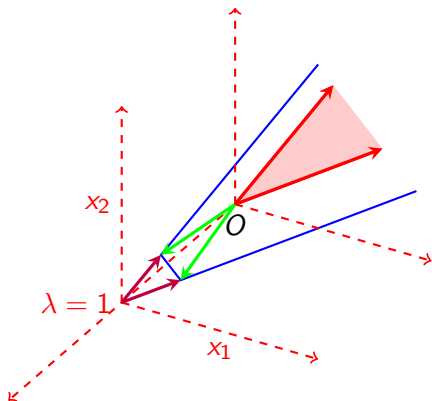


Example: $P = Q + C$

Example 10.5

Consider the following polyhedron P .

1. Green + red vectors are generators of P'
2. Red vectors have no λ component, they form the cone C
3. Green vectors have $\lambda = 1$.
4. Projecting green vectors on x_1 and x_2 plane we get purple vectors.
5. Q is the hull of the purple vectors



Farkas lemma (version I)

Theorem 10.4

Let A be a matrix and b be a vector. Then, there is a vector $x \geq 0$ such that $Ax = b$ iff

$$\text{for all } y, \quad yA \geq 0 \Rightarrow yb \geq 0.$$

Proof.

(\Rightarrow)

Let $x_0 \geq 0$ be such that $Ax_0 = b$.

Therefore, for some row vector y , $yAx_0 = yb$.

Since $x_0 \geq 0$, if $yA \geq 0$ then $yb \geq 0$.

(\Leftarrow)

Let us suppose there is no such x .

Let a_1, \dots, a_n be columns of A .

Therefore, $b \notin \text{cone}\{a_1, \dots, a_n\}$. (why?)

Due to Theorem 10.1, there is a y such that $yA \geq 0$ and $yb < 0$. □

Farkas lemma (version II)

Theorem 10.5

Let A be a matrix and b be a vector. Then, there is a vector x such that $Ax \leq b$ iff

$$\text{for all } y, \quad y \geq 0 \wedge yA = 0 \Rightarrow yb \geq 0.$$

Proof.

Consider matrix $A' = [I \ A \ -A]$.

$A'x' = b$ with $x' \geq 0$ has a solution iff $Ax \leq b$ has. (why?)

Due to theorem 10.4, the left hand side is equivalent to

$$\text{for all } y, \quad y[I \ A \ -A] \geq 0 \Rightarrow yb \geq 0.$$

Therefore, for all $y, \quad y \geq 0 \wedge yA \geq 0 \wedge -yA \geq 0 \Rightarrow yb \geq 0.$

Therefore, for all $y, \quad y \geq 0 \wedge yA = 0 \Rightarrow yb \geq 0.$



Exercise 10.4

Give the relation between solutions of $A'x' = b \wedge x' \geq 0$ and $Ax \leq b$.

Example: empty polyhedron

There is a $y \geq 0$ such that $yA = 0 \Rightarrow yb \geq 0$.

Farkas lemma (version III)

Exercise 10.5

Prove that:

Let A be a matrix and b be a vector. Then, there is a vector $x \geq 0$ such that $Ax \leq b$ iff

$$\text{for all } y, \quad y \geq 0 \wedge yA \geq 0 \Rightarrow yb \geq 0.$$

Exercise 10.6

Prove that:

Let A be a matrix and b be a vector. Then, there is a vector x such that $Ax = b$ iff

$$\text{for all } y, \quad yA = 0 \Rightarrow yb = 0.$$

Linear programming problem

Definition 10.7

Linear programming (LP) is the problem of maximizing or minimizing linear functions over a polyhedron. For example,

$$\min\{cx \mid Ax \leq b\}$$

Definition 10.8

The following is called LP-duality condition

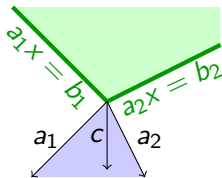
We will prove the following always holds.

$$\max\{cx \mid Ax \leq b\} = \min\{yb \mid y \geq 0 \wedge yA = c\}.$$

Example 10.6

Consider the green polyhedron with a corner. \max achieves the optima at the corner, if c is in the blue cone.

c is nonnegative combination of rows of A , i.e., y .



Duality theorem

Theorem 10.6

Let A be a matrix, and let b and c be vectors. Then,

$$\max\{cx \mid Ax \leq b\} = \min\{yb \mid y \geq 0 \wedge yA = c\}$$

provided both sets are nonempty.

Proof.

claim: \max will be less than or equal to \min

Let us suppose $Ax \leq b$, $y \geq 0$, and $yA = c$.

After multiply x in $yA = c$, we obtain $yAx = cx$.

Since $y \geq 0$ and $Ax \leq b$, $yb \geq cx$.

We need to show that the following is nonempty.

$$Ax \leq b \wedge y \geq 0 \wedge yA = c \wedge \underbrace{cx \geq yb}$$

makes min and max equal

Duality theorem (contd.)

Proof.

Writing $Ax \leq b \wedge y \geq 0 \wedge yA = c \wedge cx \geq yb$ as follows.

$$\begin{bmatrix} A & 0 \\ 0 & -I \\ 0 & A^T \\ 0 & -A^T \\ -c & b^T \end{bmatrix} \begin{bmatrix} x \\ y^T \end{bmatrix} \leq \begin{bmatrix} b \\ 0 \\ c^T \\ -c^T \\ 0 \end{bmatrix}$$

To show the above is nonempty, we apply theorem 10.5. Now we need to show that for each $u, t, v, w, \lambda \geq 0$

$$uA - \lambda c = 0 \wedge \lambda b^T + (v - w)A^T - t = 0 \Rightarrow ub + (v - w)c^T \geq 0.$$

After simplifications, we need to show that for each $u, \lambda \geq 0$ and v'

$$uA = \lambda c \wedge \lambda b^T + v'A^T \geq 0 \Rightarrow ub + v'c^T \geq 0.$$

Duality theorem (contd.)

Proof.

We need to show that for each $u, \lambda \geq 0$ and v'

$$uA = \lambda c \wedge \lambda b^T + v'A^T \geq 0 \Rightarrow ub + v'c^T \geq 0.$$

We assume left hand side and case split on number λ .

case $\lambda > 0$:

$$\begin{aligned} \text{Consider } \lambda b^T + v'A^T \geq 0 &\rightsquigarrow b^T + v'A^T/\lambda \geq 0 &\rightsquigarrow b + Av'^T/\lambda \geq 0 \\ &\rightsquigarrow ub + \lambda cv'^T/\lambda \geq 0 &\rightsquigarrow ub + cv'^T \geq 0 &\rightsquigarrow ub + v'c^T \geq 0_{(\text{why?})} \end{aligned}$$

case $\lambda = 0$:

Left hand side reduces to $uA = 0 \wedge v'A^T \geq 0$.

claim: $ub \geq 0$

There is a x_0 such that $Ax_0 \leq b$.
Therefore, $ub \geq uAx_0 = 0$.

claim: $v'c^T \geq 0$

There is a y_0 such that $y_0 \geq 0 \wedge y_0A = c$.
 $y_0^T \geq 0 \wedge v'A^Ty_0^T = v'c^T$.
Therefore, $v'c^T \geq 0$. □

Emptiness of dual space

Definition 10.9

For an LP problem $\max\{cx \mid Ax \leq b\}$, the set $\{y \mid y \geq 0 \wedge yA = c\}$ is called dual space.

Theorem 10.7

If the dual space of LP problem $\max\{cx \mid Ax \leq b\}$ is empty. Then, the maximum value is unbounded.

Proof.

Let us suppose the dual space $y \geq 0 \wedge yA = c$ is empty.

Due to theorem 10.4, there is a z such that

$$Az \geq 0 \wedge cz < 0.$$

We can use $-z$ to arbitrarily increase the value of cx . Therefore, the maximum value is unbounded. □

Farkas lemma (Affine version)

Theorem 10.8

Let the system $Ax \leq b$ is nonempty and let c be a row vector and δ be a number. Let us suppose for each x

$$Ax \leq b \Rightarrow cx \leq \delta.$$

Then there exists $\delta' \leq \delta$ such that $cx \leq \delta'$ is a nonnegative linear combination of the inequalities in $Ax \leq b$.

Proof.

Since the max is bounded, the dual space is nonempty and let the max be δ' .

Since both the spaces are nonempty and due to the duality theorem,

$$\max\{cx \mid Ax \leq b\} = \min\{yb \mid y \geq 0 \wedge yA = c\}$$

Therefore, there exists y_0 , such that $y_0 b = \delta' \wedge y_0 \geq 0 \wedge y_0 A = c$. (why?)

Therefore, $cx \leq \delta'$ is nonnegative linear combination of $Ax \leq c$. (why?)



End of Lecture 10