CS 433 Automated Reasoning 2021

Lecture 14: Theory of linear rational arithmetic (LRA)

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Topic 14.1

Theory of linear rational arithmetic



Linear rational arithmetic (LRA)

Formulas with structure $\Sigma = (\{+/2, 0, 1, ...\}, \{</2\})$ with a set of axioms Note: We have seen the axioms in the third lecture.

Example 14.1

The following formulas are in the quantifier-free fragment of the theory (QF_LRA), where x, y, and z are the rationals.

$$\triangleright \ x \ge 0 \lor y + z = 5$$

$$\blacktriangleright x < 300 \land x - z \neq 5$$

Exercise 14.1

There is no \leq in the signature. How can we use the symbol?



Proof system for QF_LRA

Due to the Farkas lemma, the following proof rule is complete for the reasoning over QF_LRA.

$$[\text{COMB}] \frac{t_1 \leq 0 \quad t_2 \leq 0}{t_1 \lambda_1 + t_2 \lambda_2 - \lambda_3 \leq 0} \lambda_1, \lambda_2, \lambda_3 \geq 0$$

Example 14.2

The following is an instance of the proof step

$$\frac{2x - y \le 1}{x + y \le 5} \frac{4y - 2x \le 6}{\lambda_1} = 1, \lambda_2 = 0.5, \lambda_3 = 1$$

Example 14.3

The following is an another instance of the proof step that derives false.

$$\frac{x + y \leq -2 \quad -x \leq 0 \quad -y \leq 1}{0 \leq -1}$$

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Theory solver for rational linear arithmetic

We will discuss the following method to find satisfiability of conjunction of linear inequalities.

Simplex

We may cover some of the following methods in the next lecture.

- Fourier-Motzkin
- Elliposid method
- Kermakar's method

We present the above methods using non-strict linear inequalities. However, they are extendable to strict inequalities, equalities, disequalities.



Topic 14.2

Simplex



Simplex was originally designed for linear optimization problems, e.g., $max\{cx|Ax \le b\}$..

A simplex variation is used to check satisfiablity, called incremental simplex.

Commentary: In fact, there are several design choices for implementing simplex. The presentation here is one version of simplex.



Incremental simplex

- Incremental simplex
 - takes atoms one by one,
 - maintains a current assignment that satisfies the atoms seen so far, and
 - ▶ after receiving a new atom $cx \leq \delta$,
 - attempts to move the assignment in the direction of -c (optimization like operation)





Incremental simplex: unsatisfiable input

Simplex may fail to reach $cx = \delta$ and the input is unsatisfiable



Exercise 14.2

Who is responsible for the unsatisfiability?



Incremental simplex as theory solver

Recall the expected interface for SMT solver:

- push(): add new atom to the simplex state.
- pop(): inexpensive operation
- unsatCore(): again inexpensive operation



Topic 14.3

Simplex - terminology



Notation

Consider the conjunction of linear inequalities in matrix form

$$Ax \leq b$$
,

where A is a $m \times n$ matrix.

By introducing fresh variables, we transform the above into

$$\begin{bmatrix} -I & A \end{bmatrix} \begin{bmatrix} s \\ x \end{bmatrix} = 0 \text{ and } s \leq b.$$

s are called slack variables. Since there is no reason to distinguish *x* and *s* in simplex, *A* will refer to $\begin{bmatrix} -I & A \end{bmatrix}$ and *x* will refer to $\begin{bmatrix} s \\ x \end{bmatrix}$.



Notation (contd.)

In general, the constraints will be denoted by

$$Ax = 0$$
 and $\bigwedge_{i=1}^{m+n} l_i \leq x_i \leq u_i.$

 l_i and u_i are $+\infty$ and $-\infty$ if there is no lower and upper bound, respectively.

- A is $m \times (m + n)$ matrix.
- Since Ax = 0 defines an *n*-dim subspace in (m + n)-dim space, if we choose values of *n* variables then we fix values of the other *m* variables.
- We will refer to *i*th column of A as the column corresponding to x_i .



Example: notation

Example 14.4

Consider: $-x + y \le -2 \land x \le 3$

We introduce slack variables s_1 and s_2 for each inequality.

In matrix form,

$$\begin{bmatrix} -1 & 0 & -1 & 1 \\ 0 & -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ -\frac{s_1}{x} \\ y \end{bmatrix} = 0 \qquad \qquad \begin{array}{c} s_1 \leq -2 \\ s_2 \leq 3 \\ s_2 \leq 3 \end{array}$$



Basic and nonbasic variables

Definition 14.1

Simplex assumes all the columns of -I (of size $m \times m$) occur in A.

- The variables corresponding to the columns are called basic variables.
- Others are called nonbasic variables.



Exercise 14.3

What are the numbers of basic and nonbasic variables ?

Example: Basic and nonbasic variables

Definition 14.2

Let B be the set of indexes for the basic variables and $NB \triangleq 1..(m + n) - B$. For $j \in B$, let k_j be a row such that $A_{k_j j} = -1$ and we may write

$$x_j = \sum_{i \in NB} a_{k_j i} x_{i_j}$$

which is called the definition of x_j .

Example 14.5

$$\left[\begin{array}{cccc} -1 & 0 & -1 & 1 \\ 0 & -1 & 1 & 0 \end{array}\right] \left[\begin{array}{c} s_1 \\ s_2 \\ x \\ y \\ \end{array}\right] = 0 \qquad \qquad s_1 \le -2 \\ s_2 \le 3$$

Currently, s_1 and s_2 are basic and x and y are nonbasic. $B = \{1, 2\}$, $NB = \{3, 4\}$, $k_1 = 1$, and $k_2 = 2$. The definition of s_1 is -x + y.

Exercise 14.4

What is the definition of the other basic variable?

Current assignment

Definition 14.3

Simplex maintains current assignment $v : x \to \mathbb{Q}$ such that

 $\blacktriangleright Av = 0,$

- nonbasic variables satisfy their bounds, and,
- consequently values for basic variables in v are fixed and v may violate a bound of at most one basic variable.



Example 14.6

$$\begin{bmatrix} -1 & 0 & -1 & 1 \\ 0 & -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} s_1 \\ \frac{s_2}{-\frac{s_1}{x}} \\ y \end{bmatrix} = 0 \qquad \qquad \begin{array}{c} \begin{array}{c} \begin{array}{c} c \\ s_1 \leq -2 \\ s_2 \leq 3 \end{array} \\ s_2 \leq 3 \end{array}$$

Initially,
$$v = \{ \underbrace{x \mapsto 0, y \mapsto 0}_{, y \mapsto 0}, s_1 \mapsto 0, s_2 \mapsto 0 \}$$

Choose values for nonbasic variables, others follow!



State

Simplex ensures the following invariant. For variable $i \in NB$.

- if x_i is unbounded then $v(x_i) = 0$ and
- otherwise $v(x_i)$ is equal to one of the existing bounds of x_i

Definition 14.4 A bound on x_i is called active if $v(x_i)$ is equal to the bound. We will mark the active bounds by *.

Definition 14.5

The NB set and bound activity defines the current state of simplex.

Example 14.7

$$\begin{bmatrix} -1 & 0 & -1 & 1 \\ 0 & -1 & 1 & 0 \\ \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ x \\ y \end{bmatrix} = 0 \qquad \begin{array}{c} s_1 \leq -2 \\ s_2 \leq 3 \\ \end{array}$$

Since all nonbasic variables have no bounds, no bound is marked active.



Topic 14.4

Simplex - pivot operation



Pivot operation

If v violates a bound of a basic variable, then simplex corrects it by pivoting.

Definition 14.6

Let us suppose x_j is basic, column j has -1 at row k, and x_i is nonbasic. A pivot operation between i and j exchanges the role between x_i and x_j , i.e., row operations until column i has a single nonzero entry -1 at row k.





Variables for pivot operations

Three variables are involved in the pivoting

- 1. the violated basic variable
- 2. nonbasic variable for pivot
- 3. basic variable for pivot

The violated basic variable does not participate in pivoting.

Commentary: The above claim is not entirely accurate. In a special case, the violated basic variable may participate in pivoting. Otherwise, the violated variable remains basic variables after pivot.



Violated basic variable

Wlog, let $1 \in B$, $k_1 = 1$, and $v(x_1)$ violates u_1 .

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We need to decrease v(x_1).
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We call v(x_1) - u_1 violation difference.
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Exercise 14.5 Write other cases that are ignored due to "wlog"



Choosing nonbasic column for pivot

Since $x_1 = \sum_{i \in NB} a_{1i}x_i$, we need to change $v(x_i)$ of some x_i such that $a_{1i}x_i$ decreases

Definition 14.7

- A column $i \in NB$ is suitable if
 - \triangleright x_i is unbounded,
 - $v(x_i) = u_i \text{ and } a_{1i} > 0, \text{ or }$

▶
$$v(x_i) = l_i \text{ and } a_{1i} < 0.$$

i is selected suitable column if *i* is the smallest suitable column.

Example 14.8

$$\begin{bmatrix} -1 & 0 & -1 & 1 \\ 0 & -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} s_1 \\ s_2 \\ x \\ y \end{bmatrix} = 0 \qquad \begin{array}{c} s_1 \leq -2 \\ s_2 \leq 3 \end{array}$$

Column 3 and 4 are suitable.

Choosing basic column for pivot I

So far: v satisfies all bounds except u_1 and $i \in NB$ is the selected suitable variable.

Since x_i appears also in the definitions of the basic variables, change in $v(x_i)$ may lead to the other violations.

Consider the following definition of $j \in B$.

$$x_j = \mathbf{a}_{k_j i} x_i + \sum_{i' \in NB - \{i\}} \mathbf{a}_{k_j i'} x_{i'},$$

If $a_{k_i i} \neq 0$, changes in x_i will change x_j .

Since we assume single violation at a time, we have $l_j \leq v(x_j) \leq u_j$.



Choosing basic column for pivot - available slack

Consider again the following definition of $j \in B$.

$$x_j = \mathbf{a}_{k_j i} x_i + \sum_{i' \in \mathbf{NB} - \{i\}} \mathbf{a}_{k_j i'} x_{i'},$$

If $a_{k_i i} > 0$, if we increase x_i it will increase x_j .

The following amount is the maximum x_i can increase without violating x_i upper bound u_j .

$$\frac{u_j - v(x_j)}{a_{k_i i}}$$

Exercise 14.6 What is the expression for maximum allowed change if $a_{k_ii} < 0$?



Choosing basic column for pivot : index that allows minimum change

Wlog, let $a_{1i} < 0$. Therefore, we need to increase $v(x_i)$.

Definition 14.8

We need to find the maximum allowed change.

$$ch := min\{rac{u_j - v(x_j)}{a_{k_j i}} | a_{k_j i} > 0 \land j \in B\} \cup \{rac{l_j - v(x_j)}{a_{k_j i}} | a_{k_j i} < 0 \land j \in B\}$$

We choose the smallest *j* for which the above min is attained.

Exercise 14.7

What are the other cases in the without loss of generality?



Example: choosing basic column for pivot

Example 14.9

We change x(selected suitable column) to reduce violation difference. Since v(y) = 0 and we are varying x, $s_1 = -x$ and $s_2 = x$.

The bounds on basic variables are $s_1 \leq -2$, and $s_2 \leq 3$.



Simplex - pivoting operation to reduce violation difference

We carry *ch* and *j* from the last slide. Wlog, $ch = \frac{u_j - v(x_j)}{a_{k_j i}}$.

Now there are three possibilities

- 1. If $ch = u_i = +\infty$, pivot between *i* and 1 and activate u_1
- 2. If $ch > (u_i l_i)$, we assign $v(x_i) = l_i$ and no pivoting
- 3. Otherwise, we apply pivoting between nonbasic *i* and basic *j*. We activate u_j bound on variable x_j .

If the violation persists, we apply further pivot operations.

Theorem 14.1

Pivoting operation never increases violation difference



Example: pivoting

Example 14.10

Our running example, s_1 is in violation, chosen nonbasic column is 3 and chosen basic column is 2

$$\left[\begin{array}{rrrrr} -1 & 0 & -1 & 1 \\ 0 & -1 & 1 & 0 \end{array}\right] \left[\begin{array}{rrrr} s_1 \\ s_2 \\ x \\ y \end{array}\right] = 0 \qquad \begin{array}{rrrrr} s_1 \leq -2 \\ s_2 \leq 3 \end{array}$$

After pivoting between 3 and 2.

Now v is satisfying.

Exercise 14.8 What is v?



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Topic 14.5

Incremental simplex



Incremental simplex and single violation assumption

Before adding next atom, simplex has a solution of atoms added so far.

New atom $cx \leq \delta$ is added in the following steps.

- A fresh slack variable *s* is introduced
- s = cx is added as a row in A and $s \le \delta$ is added in the bounds
- The new row may have non-zeros in basic columns. They are removed by row operations on the new row.
- \triangleright s is added to B, declaring it to be a basic variable.

Therefore, the current assignment can only violate the bound of s.

The above strategy is called eager pivoting. We may lazily remove the violations, without breaking the correctness.



Example: inserting a new atom

Example 14.11

Let us add $-2x - y \le -8$ in our example. We add a slack variable s_3 and a corresponding row.

$$\left[\begin{array}{cccc} -1 & 0 & 0 & -2 & -1 \\ 0 & -1 & -1 & 0 & 1 \\ 0 & 0 & 1 & -1 & 0 \end{array}\right] \left[\begin{array}{cccc} s_3 \\ s_1 \\ s_2 \\ x \\ y \end{array}\right] = 0 \qquad \begin{array}{cccc} s_3 \leq -8 \\ s_1 \leq -2 \\ s_2 \leq 3^* \end{array}$$

After removing basic variables $(\{s_1, x\})$ from the top row

$$\left[\begin{array}{ccccc} -1 & 0 & -2 & 0 & -1 \\ 0 & -1 & -1 & 0 & 1 \\ 0 & 0 & 1 & -1 & 0 \end{array} \right] \left| \begin{array}{c} s_3 \\ s_1 \\ s_2 \\ x \\ y \end{array} \right| = 0 \qquad \begin{array}{c} s_3 \leq -8 \\ s_1 \leq -2 \\ x \\ s_2 \leq 3^* \end{array}$$

Exercise 14.9

Now s₃ is violated. Pivot if possible.

Simplex - iterations

Simplex is a sequence of pivot operations

- If a state is reached without violation then v is a satisfying assignment.
- ▶ If there are no suitable columns to repair a violation then input is unsat.

Example 14.12

 s_3 is still in violation.

Now, we can not find a suitable column. Therefore, the constraints are unsat.



Example 14.13

Run simplex on $x_1 \le 5 \land 4x_1 + x_2 \le 25 \land -2x_1 - x_2 \le -25$ After push of the first atom

 $\begin{bmatrix} -1 & 1 \end{bmatrix} \begin{bmatrix} s_1 \\ x_1 \end{bmatrix} = 0 \qquad s_1 \leq 5 \qquad v = \{x_1 \mapsto 0, s_1 \mapsto 0\}$

After push of the second atom

After push of the last atom Γ_{c} , \neg

$$\begin{bmatrix} -1 & 0 & 0 & -2 & -1 \\ 0 & -1 & 0 & 4 & 1 \\ 0 & 0 & -1 & 1 & 0 \end{bmatrix} \begin{vmatrix} s_3 \\ s_2 \\ s_1 \\ x_1 \\ x_2 \end{vmatrix} = \begin{bmatrix} s_1 \le 5 \\ 0 & s_2 \le 25 \\ s_3 \le -25 \end{vmatrix} \quad v = \{ _ \mapsto 0 \$$

Exercise 14.10Finish the run



Theory solver interface pop()

If we want to remove some atom from simplex state, we

- \triangleright make the corresponding slack variable x_i basic variable and
- remove the corresponding row k_i and bound constraints on x_i

Cost: one pivot operation



Theory solver interface UnsatCore()

If input is unsat, there must be a violated basic variable x_j

- we collect the slack variables that appear in the row k_j
- ▶ the atoms corresponding to the slack variables are part of unsat core

Cost: zero.

However, we used the simplex design that excessively uses slack variables.

Commentary: Some times slack variables can be avoided. For example, input atom is equality. We can solve the constraints without introducing slack variables.

Topic 14.6

Complexity of simplex



An example of worst case Simplex

The previous example is the case of exponential number of pivots.



Worst case Simplex

For *n* dimensional problem, we can make simplex do $2^n - 1$ pivots, i.e., walking on $2^n - 1$ edges of a *n*-dimensional cuboid.

Example 14.14

Problem input for three dimensions

 $x_1 \leq 5 \wedge 4x_1 + x_2 \leq 25 \wedge 8x_1 + 4x_2 + x_3 \leq 125 \wedge -4x_1 - 2x_2 - x_3 \leq -125$





Simplex is average time linear and worst case exponential.

In practice, none of the above complexities are observed

Ellipsoid method is a polynomial time algorithm for linear constraints. In practice, simplex performs better in many classes of problems.



Exercise: notation

$$\begin{bmatrix} -1 & 0 & 0 & -2 & -1 \\ 0 & -1 & 0 & 4 & 1 \\ 0 & 0 & -1 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \end{bmatrix} = \begin{bmatrix} s_1 \le 5 \\ s_2 \le 25 \\ s_3 \le -25 \end{bmatrix} \quad v = \{ _ \mapsto 0 \}$$

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Topic 14.7

Extra slides : Incremental simplex - geometric intuition



Now we will connect the algorithm with a geometric intuition.

For ease of exposition, we will assume that all l_i s and u_i s are finite. This restriction can be easily dropped.

We will add supper script p to various objects to denote their value at pth iteration.

For example, A^p is the value of A at pth iteration.



Simplex - geometric intuition: meaning of suitable column

Let us introduce the following object in each iterations

• Let μ^p be a row vector of length 2(m + n) such that

$$\begin{bmatrix} 1 & 0 & \mu^p \\ & m-1 & \\ \end{bmatrix} \begin{bmatrix} A^p & \\ I \\ -I \end{bmatrix} = \begin{bmatrix} -1 & 0 \\ & m+n-1 \end{bmatrix}$$

$$\mu_{k}^{p} = \begin{cases} -A_{1k}^{p} & k \in NB^{p} \text{ and } u_{k} \text{ is active at } p\text{th iteration} \\ A_{1(k-(m+n))}^{p} & (k-(m+n)) \in NB^{p} \text{ and } I_{k-(m+n)} \text{ is active at } p\text{th iteration} \\ 0 & \text{otherwise} \end{cases}$$

Theorem 14.2

Let i' be the smallest index for which μ^p has a negative number and i be the selected suitable column for the next pivoting. Then,

$$i = \begin{cases} i' & i' \le m + n \\ i' - (m + n) & otherwise. \end{cases}$$

Exercise 14.11

Prove the above. ©000 CS 433 Automated Reasoning 2021

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Simplex - geometric intuition: update direction

Selection of suitable column induces the idea of update direction

Let y^p be a vector of length m + n. y^p indicates the direction of change due to pivot operation after *p*th iteration.

Let $i \in NB^p$ be the selected suitable column.

I_i is active

$$y_j^p = egin{cases} 1 & j=i, \ A_{k_j i}^p & j\in B^p \ 0 & ext{otherwise} \end{cases}$$

 \blacktriangleright u_i is active

$$y_j^p = egin{cases} -1 & j=i, \ -A_{k_ji}^p & j\in B^p \ 0 & ext{otherwise} \end{cases}$$

Exercise 14.12 Show $[-1 \ 0]y^p > 0$

Simplex - geometric intuition: limit on update

The change in direction y only violate bounds on basic variables

$$ch := \min \bigcup_{j \in 1..m} \{ \frac{u_j - v(x_j)}{y_j^p} | y_j^p > 0 \} \cup \{ \frac{l_j - v(x_j)}{-y_j^p} | y_j^p > 0 \}$$

Let j be the smallest index for which the above min is attained, which is used for pivoting. Exercise 14.13

Check the basis column j selected above is same as the pivot basis column selected earlier



Simplex - termination

Lemma 14.1 Simplex terminates.

Proof.

In every step the violation difference $(v(x_1) - u_1)$ reduces or stays same.

Since there are finitely many states, simplex terminates if $v(x_1) - u_1$ cannot stay same forever.

For that we prove that same state can not repeat in a simplex run.

Wlog, let us suppose the states of sth and tth iterations of simplex is same and there is no change in $v(x_1) - u_1$ from p to q.

Let r be the largest index column which left and reentered NB at iteration p and q respectively, where $s \le p < q \le t$.



Simplex - termination(contd.) Now Consider,

$$\begin{bmatrix} 1 & 0 & \mu^p \\ m-1 & 0 \end{bmatrix} \begin{bmatrix} A^p & I \\ -I & -I \end{bmatrix} y^q = \begin{bmatrix} -1 & 0 \end{bmatrix} y^q > 0$$

Now we will show that the above term cannot be > 0. Let us apply a different calculation on the above term.

$$\begin{bmatrix} 1 & \underbrace{0}_{m-1} & \mu^{p} \end{bmatrix} \begin{bmatrix} A^{p} \\ I \\ -I \end{bmatrix} y^{q} = \begin{bmatrix} 1 & \underbrace{0}_{m-1} & \mu^{p} \end{bmatrix} \begin{bmatrix} A^{p} y^{q} \\ Iy^{q} \\ -Iy^{q} \end{bmatrix}$$
$$= \begin{bmatrix} 1 & \underbrace{0}_{m-1} & \mu^{p} \end{bmatrix} \begin{bmatrix} 0 \\ Iy^{q} \\ -Iy^{q} \end{bmatrix} = \mu^{p} \begin{bmatrix} y^{q} \\ -y^{q} \\ -y^{q} \end{bmatrix}$$

Termination

Let
$$\hat{y}^q \triangleq \begin{bmatrix} y^q \\ -y^q \end{bmatrix}$$

Now we show every $\mu_j \hat{y}_j^p$ is non-positive.

▶
$$j \in B^p$$
 or $j - n \in B^p$ or j th bound is inactive, $\mu_j^p = 0$
▶ $j \in NB^p$ or $j - n \in NB^p$, and j th bound is active
▶ $j > r, y_i^q = 0$
▶ $j = r, u_r^p < 0$ and $y_r^q > 0(why?)$
because r is selected to leave NB^p
▶ $j > r, u_i^p \ge 0$ and $y_i^q \le 0$ (why?)



End of Lecture 14

