

# Automated Reasoning 2018

## Lecture 12: Other methods for LRA

Instructor: Ashutosh Gupta

IITB, India

Compile date: 2018-08-24

# Ideas that do not work!!

We have been seeing the success of SMT solvers and their algorithms.

Today, we will look at the methods that may (or may not) look efficient on paper but do not perform well!

This lecture is a word of caution, we may have **stuck in a local maxima of ideas!!**

## Where are going to look a few less popular methods

- ▶ Fourier-Motzkin
- ▶ Interior point methods
- ▶ Ellipsoid method
- ▶ Kermaker's method

# Topic 12.1

## Fourier-Motzkin

## Fourier-Motzkin

The algorithm proceeds by eliminating variables one by one. After eliminating all the variables, if the input reduces to  $\top$  then only the input is satisfiable.

For variable  $x$ , a conjunction of linear inequalities can be transformed into the following form, where  $x$  does not occur in the linear terms  $s_j$ ,  $t_k$ , and  $u_k$ .

$$\bigwedge_{j=1}^m s_j \leq x \wedge \bigwedge_{i=1}^{\ell} x \leq t_i \wedge \bigwedge_{k=1}^n u_k \leq 0$$



$$\bigwedge_{j=1}^m \bigwedge_{i=1}^{\ell} s_j \leq t_i \wedge \bigwedge_{k=1}^n u_k \leq 0$$

equisatisfiable but  
without  $x$

### Exercise 12.1

- Add support for equality, dis-equality, and strict inequalities
- What is the complexity?

## Example: Fourier-Motzkin

### Example 12.1

Consider:  $-x_1 + x_2 + 2x_3 \leq 0 \wedge x_1 - x_2 \leq 0 \wedge x_1 - x_3 \leq 0 \wedge 1 - x_3 \leq 0$

Suppose we eliminate  $x_1$  first. We transform the constraints into our format.

$$\underbrace{x_2 + 2x_3 \leq x_1}_{x_1 \text{ lower bounded}} \wedge \underbrace{(x_1 \leq x_2 \wedge x_1 \leq x_3)}_{x_1 \text{ upper bounded}} \wedge \underbrace{-x_3 + 1 \leq 0}_{x_1 \text{ does not occur}}$$

Eliminated constraints:  $x_2 + 2x_3 \leq x_2 \wedge x_2 + 2x_3 \leq x_3 \wedge -x_3 + 1 \leq 0$

After simplification:  $x_3 \leq 0 \wedge x_2 + x_3 \leq 0 \wedge -x_3 + 1 \leq 0$

Since  $x_2$  has no lower bound, we can drop  $x_2$  atoms:  $x_3 \leq 0 \wedge -x_3 + 1 \leq 0$

Eliminating  $x_3$ :  $1 \leq 0 \leftarrow$  false formula therefore unsat

### Exercise 12.2

How to generate the proof for unsatisfiable conjunctions

## Another example

### Example 12.2

Consider:  $-x_1 + x_2 + 2x_3 \leq 0 \wedge x_1 - x_2 \leq 0 \wedge x_1 - x_3 \leq 0 \wedge -x_1 \leq 2$

Let us transform to eliminate  $x_1$ .

$$\underbrace{x_2 + 2x_3 \leq x_1 \wedge -2 \leq x_1}_{x_1 \text{ lower bounded}} \wedge \underbrace{(x_1 \leq x_2 \wedge x_1 \leq x_3)}_{x_1 \text{ upper bounded}}$$

Eliminated constraints:  $x_2 + 2x_3 \leq x_2 \wedge x_2 + 2x_3 \leq x_3 \wedge -2 \leq x_2 \wedge -2 \leq x_3$

After simplification:  $x_3 \leq 0 \wedge x_2 + x_3 \leq 0 \wedge -2 \leq x_2 \wedge -2 \leq x_3$

Eliminating  $x_2$ :  $x_3 \leq 0 \wedge -2 \leq -x_3 \wedge -2 \leq x_3$

After simplification:  $x_3 \leq 0 \wedge -2 \leq x_3$

Eliminating  $x_3$ :  $-2 \leq 0 \leftarrow$  true formula therefore sat

### Exercise 12.3

How to generate the model for satisfiable conjunctions?

## Fourier-Motzkin in practice

Both complexity and practical performance of the algorithm are bad.

Almost never used in practice, except for some bounded simplifications.



## Topic 12.2

### Ellipsoid method

# Ellipsoid method : a Soviet scare

Khachian found the first polynomial time method for linear programming.

ARCHIVES | 1979

## *A Soviet Discovery Rocks World of Mathematics*

By MALCOLM W. BROWNE NOV. 7, 1979



A surprise discovery by an obscure Soviet mathematician has rocked the world of mathematics and computer analysis, and experts have begun exploring its practical applications.

<http://www.nytimes.com/1979/11/07/archives/a-soviet-discovery-rocks-world-of-mathematics-russians-surprise.html>

# Ellipsoid method

We want to check satisfiability of

$$Ax \leq b.$$

We assume that space  $Ax \leq b$  is bounded and full dimensional.

## Coefficient bounded solution existence

Size are define as follows

- ▶ For a number  $r$ ,  $size(r)$  is  $\log_2(r)$
- ▶ For a row/vector  $c$ ,  $c = n + \sum_i size(c_i)$
- ▶ For a matrix  $A$ ,  $size(A) = mn + \sum_{ij} size(A_{ij})$

### Theorem 12.1

*If solution exists of  $Ax \leq b$ , then there is a solution with size less than  $4n^2\phi$ , where  $\phi$  is the maximum row size.*

### Proof.

Every vertex is a solution of  $n$  equalities  $A'x = b'$  from  $Ax \leq b$ .

The vertex is  $x = A'^{-1}b'$ , which depends on the size of determinant of  $A'$ .

Since determinant sums multiples of  $n$  numbers, its size is bounded by  $2n\phi$ .

Therefore,  $size(x_i) \leq 4n\phi$ . Therefore,  $size(x) \leq 4n^2\phi$ . □

# Minimum volume condition

## Theorem 12.2

*If  $Ax \leq b$  is satisfiable, volume of  $Ax \leq b$  is bigger than  $2^{-2n(4n^2\phi)}$*

## Proof sketch.

Since

- ▶ all numbers have finite precision,
- ▶ the  $Ax \leq b$  is sat, and
- ▶ the  $Ax \leq b$  is full-rank,

there is a smallest volume convex hull that  $Ax \leq b$  can represent.

Therefore, there is a lower bound. □

# Idea!

- ▶ We know that if there is a solution, it is in the finite space.
- ▶ The finite space can be divided in finite granularity!!
- ▶ If we can iteratively divide the space, we may have an efficient algorithm.

We use ellipses to describe and split the finite space.

# Positive definite matrix

## Definition 12.1

A symmetric matrix is *positive definite* if all its eigenvalues are positive.

## Theorem 12.3

The following statements are equivalent

1.  $D$  is positive definite
2.  $D = B^T B$  for some non-singular  $B$
3. For each  $x$ ,  $x^T D x > 0$

## Proof.

(1) $\Rightarrow$ (2)

- ▶  $D$  can be diagonalized, i.e.,  $D = P^T D' P$  where  $D'$  is a diagonal matrix
- ▶ Since all eigenvalues are positive we can split  $D' = D'' D''$
- ▶ Therefore,  $B = P^T D''$  □

## Exercise 12.4

a. Prove (2) $\Rightarrow$ (3)      b. Prove (3) $\Rightarrow$ (1)

# Representing ellipses

## Definition 12.2

The following defines interior of an ellipse

$$\text{ell}(z, D) := \{x \mid (x - z)^T D^{-1} (x - z) \leq 1\}$$

where  $D$  is a  $n \times n$  **positive definite** matrix.

- ▶  $z$  is the center of the ellipse
- ▶  $D$  defines the direction and length of axes

## Example 12.3

2-D unit ball is

$$\text{ell}\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \mathbf{I}\right) = \{x \mid x^T \mathbf{I} x \leq 1\} = \{x \mid |x|^2 \leq 1\}$$

Shorthand for ball  $\text{ball}(z, r) := \text{ell}(z, r^2 \mathbf{I})$

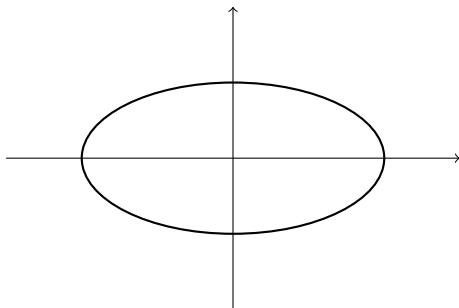


## Ellipse example: stretched

### Example 12.4

*Ellipse  $x_1^2 + 4x_2^2 \leq 4$  will be encoded as follows*

$$\text{ell}\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}\right) = \left\{x \mid x^T \begin{bmatrix} 1/4 & 0 \\ 0 & 1 \end{bmatrix} x \leq 1\right\}$$

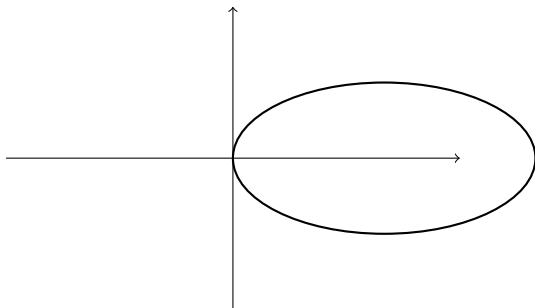


## Ellipse example: shifted

### Example 12.5

*Ellipse  $x_1^2 - 4x_1 + 4 + 4x_2^2 \leq 4$  will be encoded as follows*

$$\text{ell}\left(\begin{bmatrix} 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 4 & 0 \\ 0 & 1 \end{bmatrix}\right) = \left\{x \mid \left(x - \begin{bmatrix} 2 \\ 0 \end{bmatrix}\right)^T \begin{bmatrix} 1/4 & 0 \\ 0 & 1 \end{bmatrix} \left(x - \begin{bmatrix} 2 \\ 0 \end{bmatrix}\right) \leq 1\right\}$$

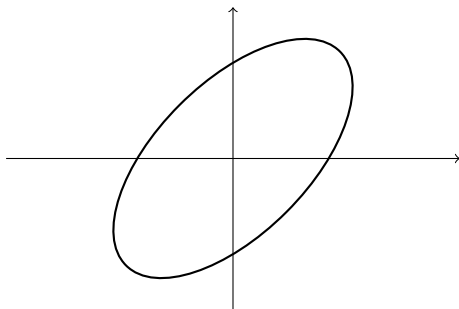


## Ellipse example: rotated

### Example 12.6

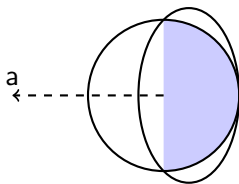
*Ellipse  $5x_1^2 + 5x_2^2 - 6x_1x_2 \leq 8$  will be encoded as follows*

$$\text{ell}([0, 0], \begin{bmatrix} \frac{5}{8} & \frac{-3}{8} \\ \frac{-3}{8} & \frac{5}{8} \end{bmatrix}^{-1}) = \{x | x^T \begin{bmatrix} \frac{5}{8} & \frac{-3}{8} \\ \frac{-3}{8} & \frac{5}{8} \end{bmatrix} x \leq 1\}$$



## Smallest covering ellipse

What is the smallest ellipse in area that covers a half circle?



Let  $a$  be the unit row vector that defines the half circle, i.e.,  
 $Ball(0, 1) \cap ax \leq 0$ .

The **smallest ellipse** is

$$ell\left( -\frac{a^T}{n+1}, \frac{n^2}{n^2-1} \left( I - \frac{2a^T a}{n+1} \right) \right)$$

## Volume decays exponentially

One axis of the ball is shrunk by  $\frac{n}{n+1}$ .

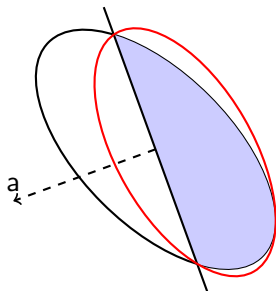
And the other  $n - 1$  axes are expanded by  $\frac{n^2}{n^2 - 1}^{1/2}$ . (why?)

Therefore the volume is changed by the factor of  $\frac{n}{n+1} \frac{n^2}{n^2 - 1}^{(n-1)/2}$

$$\frac{n}{n+1} \frac{n^2}{n^2 - 1}^{(n-1)/2} < e^{-1/(2n+2)}. \text{ (why?)}$$

## In more general form

Consider the following ellipse  $ell(z, D)$  and direction  $a$



The smaller COVERING ELLIPSE  $(z, D, a) :=$

$$ell\left( z - \frac{Da^T}{(n+1)\sqrt{aDa^T}}, \frac{n^2}{n^2-1} \left( D - \frac{2Da^T aD}{(n+1)\sqrt{aDa^T}} \right) \right)$$

Since all ellipses are linear transformations of unit ball, the exponential volume reduction still holds.

# Ellipsoid method

Input  $Ax \leq 0$ .

We know if  $(Ax \leq 0)$  is sat,  
 $(Ax \leq 0) \subseteq \text{ball}(0, 2^{4n^2\phi})$ .

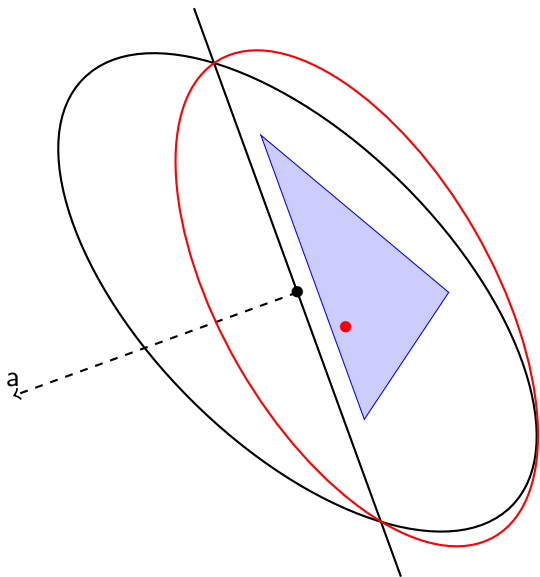
Let us suppose we have initial ellipse  $\text{ell}(z, D) := \text{ball}(0, 2^{4n^2\phi})$ .

1. if  $z$  satisfies  $Ax \leq b$ , **return**  $z$
2. Otherwise, find inequality  $ax \leq \delta$  in  $Ax \leq b$  such that  $az > \delta$
3.  $z, D := \text{COVERINGELLIPSE}(z, D, a)$
4. If volume of  $\text{ell}(z, D)$  is **too small**, **return** unsatisfiable
5. goto 1

## Exercise 12.5

- a. Why  $ax \leq \delta$  exists at 2?
- b. Why smaller ellipses will continue to contain  $Ax \leq b$ ?

# Ellipsoid method illustration





# Ellipsoid method is polynomial

## Theorem 12.4

*Ellipsoid method runs less than  $16n^2(4n^2\phi)$  iterations.*

## Proof.

1. Initial ellipse has volume less than  $(2 \times 2^{4n^2\phi})^n$
2. Volume threshold is  $2^{-2n(4n^2\phi)}$
3. Ellipse sizes decreases by the factor of  $e^{-1/2(n+1)}$



## Exercise 12.6

*Prove that within the  $16n^2(4n^2\phi)$  iterations the ellipse volume will reduce below the threshold*

## Inefficient method

- ▶ Number of iterations depends on the size of the numbers
- ▶ Square root needs to be computed, i.e., high precision computation (can be avoided!)
- ▶ Experiments show that it can not compete with simplex.

## Topic 12.3

### Karmakar's method

# Karmakar's method: west strikes back

Karmakar fixed some of the problems in ellipsoid method

ARCHIVES | 1984

## ***BREAKTHROUGH IN PROBLEM SOLVING***

By JAMES GLEICK



A 28-year-old mathematician at A.T.&T. Bell Laboratories has made a startling theoretical breakthrough in the solving of systems of equations that often grow too vast and complex for the most powerful computers.

The discovery, which is to be formally published next month, is already circulating rapidly through the mathematical world. It has also set off a deluge of inquiries from brokerage houses, oil companies and airlines, industries with millions of dollars at stake in problems known as linear programming.

<http://www.nytimes.com/1984/11/19/us/breakthrough-in-problem-solving.html>

# Efficacy of Karmakar's method

- ▶ There are claims that it is far more efficient
- ▶ No large scale study to demonstrate (as far as I know!!)
- ▶ All SMT solvers still implement simplex

End of Lecture 12