# Complexity

### LP Feasibility Problem (LP feasibility)

- Given  $A \in \mathbb{Z}^{m \times n}$ ,  $b \in \mathbb{Z}^m$ . Does there exist  $x \in \mathbb{R}$  with  $Ax = b, x \ge 0$ ?
- Note that allowing A, b to contain rational numbers does not make a difference, as we can multiply every number by a suitable large constant so that everything becomes integral but the feasible region does not change.

### Is this problem in NP or even in P?

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- In the following we sometimes refer to  $L := L(\lceil A \mid b \rceil)$  as the input size (even though the real input size is something in  $\Theta(L([A|b]))).$
- In order to show that LP-decision is in NP we show that if there is a solution x then there exists a small solution for which feasibility can be verified in polynomial time (polynomial in L([A|b])).

# The Bit Model

### Input size

• The number of bits to represent a number  $a \in \mathbb{Z}$  is

### $\lceil \log_2(|a|) \rceil + 1$

• Let for an  $m \times n$  matrix M, L(M) denote the number of bits required to encode all the numbers in *M*.

$$L(M) := \sum_{i,j} \lceil \log_2(|m_{ij}|) + 1 \rceil$$

- In the following we assume that input matrices are encoded in a standard way, where each number is encoded in binary and then suitable separators are added in order to separate distinct number from each other.
- Then the input length is  $\Theta(L([A|b]))$ .

Suppose that Ax = b;  $x \ge 0$  is feasible.

Then there exists a basic feasible solution. This means a set B of basic variables such that

 $x_B = A_B^{-1}b$ 

and all other entries in x are 0.

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# Size of a Basic Feasible Solution

### Lemma 2

Let  $M \in \mathbb{Z}^{m \times m}$  be an invertable matrix and let  $b \in \mathbb{Z}^m$ . Further define  $L' = L([M | b]) + n \log_2 n$ . Then a solution to Mx = b has rational components  $x_j$  of the form  $\frac{D_j}{D}$ , where  $|D_j| \le 2^{L'}$  and  $|D| \le 2^{L'}$ .

**Proof:** Cramers rules says that we can compute  $x_j$  as

 $x_j = \frac{\det(M_j)}{\det(M)}$ 

where  $M_j$  is the matrix obtained from M by replacing the j-th column by the vector b.

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This means if Ax = b,  $x \ge 0$  is feasible we only need to consider vectors x where an entry  $x_j$  can be represented by a rational number with encoding length polynomial in the input length L.

Hence, the x that we have to guess is of length polynomial in the input-length L.

For a given vector x of polynomial length we can check for feasibility in polynomial time.

Hence, LP feasibility is in NP.

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# **Bounding the Determinant**

Let  $X = A_B$ . Then

$$|\det(X)| = \left| \sum_{\pi \in S_n} \operatorname{sgn}(\pi) \prod_{1 \le i \le n} X_{i\pi(i)} \right|$$
  
$$\leq \sum_{\pi \in S_n} \prod_{1 \le i \le n} |X_{i\pi(i)}|$$
  
$$\leq n! \cdot 2^{L([A|b])} \le n^n 2^L \le 2^{L'} .$$

Analogously for  $det(M_i)$ .

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# **Reducing LP-solving to LP decision.**

Given an LP max{ $c^T x | Ax = b; x \ge 0$ } do a binary search for the optimum solution

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(Add constraint  $c^T x - \delta = M$ ;  $\delta \ge 0$  or  $(c^T x \ge M)$ . Then checking for feasibility shows whether optimum solution is larger or smaller than M).

If the LP is feasible then the binary search finishes in at most

$$\log_2\left(\frac{2n2^{2L'}}{1/2^{L'}}\right) = \mathcal{O}(L') ,$$

as the range of the search is at most  $-n2^{2L'}, \ldots, n2^{2L'}$  and the distance between two adjacent values is at least  $\frac{1}{\det(A)} \ge \frac{1}{2^{L'}}$ .

Here we use  $L' = L([A | b | c]) + n \log_2 n$  (it also includes the encoding size of *c*).



Let  $M_{\text{max}} = n2^{2L'}$  be an upper bound on the objective value of a basic feasible solution.

We can add a constraint  $c^T x \ge M_{max} + 1$  and check for feasibility.

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# **Ellipsoid Method**

- Let *K* be a convex set.
- Maintain ellipsoid E that is guaranteed to contain K provided that K is non-empty.
- If center  $z \in K$  STOP.
- Otw. find a hyperplane separating *K* from *z* (e.g. a violated constraint in the LP).
   Shift hyperplane to contain

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- Shift hyperplane to contain node z. H denotes halfspace that contains K.
- Compute (smallest) ellipsoid E' that contains  $K \cap H$ .
- REPEAT

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### Issues/Questions:

- How do you choose the first Ellipsoid? What is its volume?
- ▶ What if the polytop *K* is unbounded?
- How do you measure progress? By how much does the volume decrease in each iteration?
- When can you stop? What is the minimum volume of a non-empty polytop?

### **Definition 3**

A mapping  $f : \mathbb{R}^n \to \mathbb{R}^n$  with f(x) = Lx + t, where *L* is an invertible matrix is called an affine transformation.

**Definition 4** 

A ball in  $\mathbb{R}^n$  with center *c* and radius *r* is given by

$$B(c,r) = \{x \mid (x-c)^{T} (x-c) \le r^{2}\}\$$
$$= \{x \mid \sum_{i} (x-c)_{i}^{2} / r^{2} \le 1\}\$$

B(0,1) is called the unit ball.

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# How to Compute the New Ellipsoid

• Use  $f^{-1}$  (recall that f = Lx + t is the affine transformation of the unit ball) to rotate/distort the ellipsoid (back) into the unit ball.

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- Use a rotation  $R^{-1}$  to rotate the unit ball such that the normal vector of the halfspace is parallel to  $e_1$ .
- Compute the new center  $\hat{c}'$  and the new matrix  $\hat{O}'$  for this simplified setting.
- Use the transformations R and f to get the new center c' and the new matrix O'for the original ellipsoid *E*.



### **Definition 5**

An affine transformation of the unit ball is called an ellipsoid.

From f(x) = Lx + t follows  $x = L^{-1}(f(x) - t)$ .

$$f(B(0,1)) = \{f(x) \mid x \in B(0,1)\}$$
  
=  $\{y \in \mathbb{R}^n \mid L^{-1}(y-t) \in B(0,1)\}$   
=  $\{y \in \mathbb{R}^n \mid (y-t)^T L^{-1} L^{-1}(y-t) \le 1\}$   
=  $\{y \in \mathbb{R}^n \mid (y-t)^T Q^{-1}(y-t) \le 1\}$ 

where  $Q = LL^T$  is an invertible matrix.

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# **The Easy Case**

- The obtain the matrix  $\hat{Q'}^{-1}$  for our ellipsoid  $\hat{E'}$  note that  $\hat{E'}$  is axis-parallel.
- Let a denote the radius along the x<sub>1</sub>-axis and let b denote the (common) radius for the other axes.
- ► The matrix

 $\hat{L}' = \begin{pmatrix} a & 0 & \dots & 0 \\ 0 & b & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & b \end{pmatrix}$ 

maps the unit ball (via function  $\hat{f}'(x) = \hat{L}'x$ ) to an axis-parallel ellipsoid with radius a in direction  $x_1$  and b in all other directions.

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# **The Easy Case** • As $\hat{Q}' = \hat{L}' \hat{L}'^{t}$ the matrix $\hat{Q}'^{-1}$ is of the form $\hat{Q}'^{-1} = \begin{pmatrix} \frac{1}{a^{2}} & 0 & \cdots & 0\\ 0 & \frac{1}{b^{2}} & \ddots & \vdots\\ \vdots & \ddots & \ddots & 0\\ 0 & \cdots & 0 & \frac{1}{b^{2}} \end{pmatrix}$

The Easy Case
For $i \neq 1$ the equation $(e_i - \hat{c}')^T \hat{Q}'^{-1} (e_i - \hat{c}') = 1$ gives
$\begin{pmatrix} -t \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}^{T} \cdot \begin{pmatrix} \frac{1}{a^{2}} & 0 & \dots & 0 \\ 0 & \frac{1}{b^{2}} & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \frac{1}{b^{2}} \end{pmatrix} \cdot \begin{pmatrix} -t \\ 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} = 1$ This gives $\frac{t^{2}}{a^{2}} + \frac{1}{b^{2}} = 1$ , and hence $\frac{1}{b^{2}} = 1 - \frac{t^{2}}{a^{2}} = 1 - \frac{t^{2}}{(1-t)^{2}} = \frac{1-2t}{(1-t)^{2}}$
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# **The Easy Case**

• We want to choose t such that the volume of  $\hat{E}'$  is minimal.

 $\operatorname{vol}(\hat{E}') = \operatorname{vol}(B(0,1)) \cdot |\operatorname{det}(\hat{L}')| ,$ 

where 
$$\hat{Q}' = \hat{L}' \hat{L}'^T$$
.

We have

$$\hat{L'}^{-1} = \begin{pmatrix} \frac{1}{a} & 0 & \dots & 0 \\ 0 & \frac{1}{b} & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \frac{1}{b} \end{pmatrix} \text{ and } \hat{L'} = \begin{pmatrix} a & 0 & \dots & 0 \\ 0 & b & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & b \end{pmatrix}$$

Note that a and b in the above equations depend on t, by the previous equations.

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The Easy Case	
$\operatorname{vol}(\hat{E}') = \operatorname{vol}(B(0,1)) \cdot  \operatorname{det}(\hat{L}') $ $= \operatorname{vol}(B(0,1)) \cdot ab^{n-1}$	
$= \operatorname{vol}(B(0,1)) \cdot (1-t) \cdot \left(\frac{1-t}{\sqrt{1-2t}}\right)^{n-1}$ $= \operatorname{vol}(B(0,1)) \cdot \frac{(1-t)^n}{(\sqrt{1-2t})^{n-1}}$	
$(\sqrt{1-2t})^{n-1}$	
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# The Easy Case • We obtain the minimum for $t = \frac{1}{n+1}$ . • For this value we obtain $a = 1 - t = \frac{n}{n+1}$ and $b = \frac{1-t}{\sqrt{1-2t}} = \frac{n}{\sqrt{n^2-1}}$ To see the equation for *b*, observe that $b^2 = \frac{(1-t)^2}{1-2t} = \frac{(1-\frac{1}{n+1})^2}{1-\frac{2}{n+1}} = \frac{(\frac{n}{n+1})^2}{\frac{n-1}{n+1}} = \frac{n^2}{n^2-1}$

# **The Easy Case**

Let  $\gamma_n = \frac{\operatorname{vol}(\hat{E}')}{\operatorname{vol}(B(0,1))} = ab^{n-1}$  be the ratio by which the volume changes:

$$y_n^2 = \left(\frac{n}{n+1}\right)^2 \left(\frac{n^2}{n^2-1}\right)^{n-1}$$
  
=  $\left(1 - \frac{1}{n+1}\right)^2 \left(1 + \frac{1}{(n-1)(n+1)}\right)^{n-1}$   
 $\leq e^{-2\frac{1}{n+1}} \cdot e^{\frac{1}{n+1}}$   
=  $e^{-\frac{1}{n+1}}$ 

where we used  $(1 + x)^a \le e^{ax}$  for  $x \in \mathbb{R}$  and a > 0.

This gives  $\gamma_n \leq e^{-\frac{1}{2(n+1)}}$ .

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Our progress is the same:  $e^{-\frac{1}{2(n+1)}} \leq \frac{\operatorname{vol}(\hat{E}')}{\operatorname{vol}(B(0,1))} = \frac{\operatorname{vol}(\hat{E}')}{\operatorname{vol}(\hat{E})} = \frac{\operatorname{vol}(R(\hat{E}'))}{\operatorname{vol}(R(\hat{E}))}$   $= \frac{\operatorname{vol}(\bar{E}')}{\operatorname{vol}(\bar{E})} = \frac{\operatorname{vol}(f(\bar{E}'))}{\operatorname{vol}(f(\bar{E}))} = \frac{\operatorname{vol}(E')}{\operatorname{vol}(E)}$ Here it is important that mapping a set with affine function f(x) = Lx + t changes the volume by factor det(L).

# How to Compute the New Ellipsoid

- Use  $f^{-1}$  (recall that f = Lx + t is the affine transformation of the unit ball) to rotate/distort the ellipsoid (back) into the unit ball.
- ► Use a rotation  $R^{-1}$  to rotate the unit ball such that the normal vector of the halfspace is parallel to  $e_1$ .
- Compute the new center ĉ' and the new matrix Q' for this simplified setting.
- Use the transformations *R* and *f* to get the new center *c'* and the new matrix *Q'*  for the original ellipsoid *E*.

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# The Ellipsoid Algorithm

### How to Compute The New Parameters?

The transformation function of the (old) ellipsoid: f(x) = Lx + c;

The halfspace to be intersected:  $H = \{x \mid a^T(x - c) \le 0\};\$ 

$$f^{-1}(H) = \{f^{-1}(x) \mid a^{T}(x-c) \le 0\}$$
  
=  $\{f^{-1}(f(y)) \mid a^{T}(f(y)-c) \le 0\}$   
=  $\{y \mid a^{T}(f(y)-c) \le 0\}$   
=  $\{y \mid a^{T}(Ly+c-c) \le 0\}$   
=  $\{y \mid (a^{T}L)y \le 0\}$ 

This means  $\bar{a} = L^T a$ .

EADS II © Harald Räcke  $\hat{E}' \ \bar{E}'$ 

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After rotating back (applying  $R^{-1}$ ) the normal vector of the halfspace points in negative  $x_1$ -direction. Hence,

$$R^{-1} \Big( \frac{L^T a}{\|L^T a\|} \Big) = -e_1 \quad \Rightarrow \quad -\frac{L^T a}{\|L^T a\|} = R \cdot e_1$$

Hence,

$$\bar{c}' = R \cdot \hat{c}' = R \cdot \frac{1}{n+1}e_1 = -\frac{1}{n+1}\frac{L^T a}{\|L^T a\|}$$

$$c' = f(\bar{c}') = L \cdot \bar{c}' + c$$
$$= -\frac{1}{n+1}L\frac{L^{T}a}{\|L^{T}a\|} + c$$
$$= c - \frac{1}{n+1}\frac{Qa}{\sqrt{a^{T}Qa}}$$

Recall that

$$\hat{Q}' = \begin{pmatrix} a^2 & 0 & \dots & 0 \\ 0 & b^2 & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & b^2 \end{pmatrix}$$

This gives

$$\hat{Q}' = \frac{n^2}{n^2 - 1} \left( I - \frac{2}{n+1} e_1 e_1^T \right)$$

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because for a = n/n+1 and  $b = n/\sqrt{n^2-1}$ 

$$b^{2} - b^{2} \frac{2}{n+1} = \frac{n^{2}}{n^{2} - 1} - \frac{2n^{2}}{(n-1)(n+1)^{2}}$$
$$= \frac{n^{2}(n+1) - 2n^{2}}{(n-1)(n+1)^{2}} = \frac{n^{2}(n-1)}{(n-1)(n+1)^{2}} = a^{2}$$

For computing the matrix Q' of the new ellipsoid we assume in the following that  $\hat{E}', \bar{E}'$  and E' refer to the ellipsoids centered in the origin.



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	$\begin{split} \bar{E}' &= R(\hat{E}') \\ &= \{R(x) \mid x^T \hat{Q}'^{-1} x \le 1\} \\ &= \{y \mid (R^{-1}y)^T \hat{Q}'^{-1} R^{-1} y \le 1\} \\ &= \{y \mid y^T (R^T)^{-1} \hat{Q}'^{-1} R^{-1} y \le 1\} \\ &= \{y \mid y^T (\underbrace{R \hat{Q}' R^T}_{\hat{Q}'})^{-1} y \le 1\} \end{split}$	
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Hence,

$$\begin{split} \bar{Q}' &= R\hat{Q}'R^T \\ &= R \cdot \frac{n^2}{n^2 - 1} \left( I - \frac{2}{n+1} e_1 e_1^T \right) \cdot R^T \\ &= \frac{n^2}{n^2 - 1} \left( R \cdot R^T - \frac{2}{n+1} (Re_1) (Re_1)^T \right) \\ &= \frac{n^2}{n^2 - 1} \left( I - \frac{2}{n+1} \frac{L^T a a^T L}{\|L^T a\|^2} \right) \end{split}$$

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$$E' = L(\bar{E}')$$

$$= \{L(x) \mid x^T \bar{Q}'^{-1} x \le 1\}$$

$$= \{y \mid (L^{-1}y)^T \bar{Q}'^{-1} L^{-1} y \le 1\}$$

$$= \{y \mid y^T (L\bar{Q}' L^T)^{-1} \bar{Q}'^{-1} L^{-1} y \le 1\}$$

$$= \{y \mid y^T (\underline{L}\bar{Q}' L^T)^{-1} y \le 1\}$$
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Algori	thm 1 ellipsoid-algorithm	
1: inp	<b>ut:</b> point $c \in \mathbb{R}^n$ , convex set $K \subseteq \mathbb{R}^n$	
2: <b>out</b>	t <b>put:</b> point $x \in K$ or " $K$ is empty"	
3: <b>Q</b> ◄	- ???	
4: <b>rep</b>	eat	
5:	if $c \in K$ then return $c$	
6:	else	
7:	choose a violated hyperplane <i>a</i>	
8:	$c \leftarrow c - \frac{1}{n+1} \frac{Qa}{\sqrt{a^T Qa}}$	
9:	$Q \leftarrow \frac{n^2}{n^2 - 1} \Big( Q - \frac{2}{n+1} \frac{Qaa^T Q}{a^T Qa} \Big)$	
10:	endif	
11: <b>un</b>	til ???	

# **Repeat: Size of basic solutions**

### Lemma 7

Let  $P = \{x \in \mathbb{R}^n \mid Ax \le b\}$  be a bounded polyhedron. Let  $\langle a_{\max} \rangle$ be the maximum encoding length of an entry in A, b. Then every entry  $x_j$  in a basic solution fulfills  $|x_j| = \frac{D_j}{D}$  with  $D_j, D \le 2^{2n\langle a_{\max} \rangle + 2n \log_2 n}$ .

Note that here we have  $P = \{x \mid Ax \le b\}$ . The previous lemmas we had about the size of feasible solutions were slightly different as they were for different polytopes.

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### How do we find the first ellipsoid?

For feasibility checking we can assume that the polytop P is bounded; it is sufficient to consider basic solutions.

Every entry  $x_i$  in a basic solution fulfills  $|x_i| \le \delta$ .

Hence, *P* is contained in the cube  $-\delta \le x_i \le \delta$ .

A vector in this cube has at most distance  $R := \sqrt{n}\delta$  from the origin.

Starting with the ball  $E_0 := B(0, R)$  ensures that P is completely contained in the initial ellipsoid. This ellipsoid has volume at most  $R^n B(0, 1) \le (n\delta)^n B(0, 1)$ .

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### **Repeat: Size of basic solutions**

Proof:

Let  $\bar{A} = \begin{bmatrix} A & -A \\ -A & A \end{bmatrix}$ ,  $\bar{b} = \begin{pmatrix} b \\ -b \end{pmatrix}$ , be the matrix and right-hand vector after transforming the system to standard form.

The determinant of the matrices  $\bar{A}_B$  and  $\bar{M}_j$  (matrix obt. when replacing the *j*-th column of  $\bar{A}_B$  by  $\bar{b}$ ) can become at most

 $\det(\bar{A}_B), \det(\bar{M}_j) \le \|\vec{\ell}_{\max}\|^{2n}$  $\le (\sqrt{2n} \cdot 2^{\langle a_{\max} \rangle})^{2n} \le 2^{2n \langle a_{\max} \rangle + 2n \log_2 n} ,$ 

where  $\vec{\ell}_{max}$  is the longest column-vector that can be obtained after deleting all but 2n rows and columns from  $\bar{A}$ .

This holds because columns from  $I_m$  selected when going from  $\overline{A}$  to  $\overline{A}_B$  do not increase the determinant. Only the at most 2n columns from matrices A and -A that  $\overline{A}$  consists of contribute.

### When can we terminate?

Let  $P := \{x \mid Ax \leq b\}$  with  $A \in \mathbb{Z}$  and  $b \in \mathbb{Z}$  be a bounded polytop. Let  $\langle a_{\max} \rangle$  be the encoding length of the largest entry in A or b.

Consider the following polyhedron

$$P_{\lambda} := \left\{ x \mid Ax \leq b + rac{1}{\lambda} \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix} 
ight\} \;,$$

where  $\lambda = \delta^2 + 1$ .

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In the following we use  $\delta := 2^{2n\langle a_{\max} \rangle + 2n \log_2 n}$ .



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Let  $\bar{A} = \begin{bmatrix} A & -A \\ -A & A \end{bmatrix}$ , and  $\bar{b} = \begin{pmatrix} b \\ -b \end{pmatrix}$ .

 $\bar{P}_{\lambda}$  feasible implies that there is a basic feasible solution represented by

 $x_B = \bar{A}_B^{-1}\bar{b} + \frac{1}{\lambda}\bar{A}_B^{-1} \begin{pmatrix} 1\\ \vdots\\ 1 \end{pmatrix}$ 

(The other *x*-values are zero)

The only reason that this basic feasible solution is not feasible for  $\bar{P}$  is that one of the basic variables becomes negative.

Hence, there exists i with

$$(\bar{A}_B^{-1}\bar{b})_i < 0 \leq (\bar{A}_B^{-1}\bar{b})_i + \frac{1}{\lambda}(\bar{A}_B^{-1}\vec{1})_i$$

⇒:

Consider the polyhedrons

$$\bar{P} = \left\{ x \mid \begin{bmatrix} A & -A \\ -A & A \end{bmatrix} x = \begin{pmatrix} b \\ -b \end{pmatrix}; x \ge 0 \right\}$$

and

$$\bar{P}_{\lambda} = \left\{ x \mid \begin{bmatrix} A & -A \\ -A & A \end{bmatrix} x = \begin{pmatrix} b \\ -b \end{pmatrix} + \frac{1}{\lambda} \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}; x \ge 0 \right\}$$

P is feasible if and only if  $\bar{P}$  is feasible, and  $P_{\lambda}$  feasible if and only if  $\bar{P}_{\lambda}$  feasible.

 $\bar{P}_{\lambda}$  is bounded since  $P_{\lambda}$  and P are bounded.

By Cramers rule we get

$$(\bar{A}_B^{-1}\bar{b})_i < 0 \implies (\bar{A}_B^{-1}\bar{b})_i \le -\frac{1}{\det(\bar{A}_B)_i}$$

and

 $(\bar{A}_B^{-1}\vec{1})_i \leq \det(\bar{M}_j)$  ,

where  $\bar{M}_j$  is obtained by replacing the *j*-th column of  $\bar{A}_B$  by  $\vec{1}$ .

However, we showed that the determinants of  $\bar{A}_B$  and  $\bar{M}_j$  can become at most  $\delta.$ 

Since, we chose  $\lambda = \delta^2 + 1$  this gives a contradiction.

### Lemma 9

If  $P_{\lambda}$  is feasible then it contains a ball of radius  $r := 1/\delta^3$ . This has a volume of at least  $r^n \operatorname{vol}(B(0,1)) = \frac{1}{\delta^{3n}} \operatorname{vol}(B(0,1))$ .

### Proof:

If  $P_{\lambda}$  feasible then also P. Let x be feasible for P. This means  $Ax \leq b$ .

Let 
$$\vec{\ell}$$
 with  $\|\vec{\ell}\| \leq r$ . Then

 $(A(x+\vec{\ell}))_i = (Ax)_i + (A\vec{\ell})_i \le b_i + \vec{a}_i^T \vec{\ell}$  $\le b_i + \|\vec{a}_i\| \cdot \|\vec{\ell}\| \le b_i + \sqrt{n} \cdot 2^{\langle a_{\max} \rangle} \cdot r$  $\le b_i + \frac{\sqrt{n} \cdot 2^{\langle a_{\max} \rangle}}{\delta^3} \le b_i + \frac{1}{\delta^2 + 1} \le b_i + \frac{1}{\lambda}$ 

Hence,  $x + \vec{\ell}$  is feasible for  $P_{\lambda}$  which proves the lemma.

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Algorithm 1 ellipsoid-algorithm
 1: input: point c \in \mathbb{R}^n, convex set K \subseteq \mathbb{R}^n, radii R and r
              with K \subseteq B(c, R), and B(x, r) \subseteq K for some x
 2:
 3: output: point x \in K or "K is empty"
 4: Q \leftarrow \operatorname{diag}(R^2, \dots, R^2) // \text{ i.e., } L = \operatorname{diag}(R, \dots, R)
 5: repeat
            if c \in K then return c
 6:
           else
 7:
                  choose a violated hyperplane a
 8:
                 c \leftarrow c - \frac{1}{n+1} \frac{Qa}{\sqrt{a^T Qa}}
 9:
                  Q \leftarrow \frac{n^2}{n^2 - 1} \left( Q - \frac{2}{n+1} \frac{Qaa^T Q}{a^T \Omega a} \right)
10:
            endif
11:
12: until det(Q) \leq r^{2n} // i.e., det(L) \leq r^{n}
13: return "K is empty"
```

How many iterations do we need until the volume becomes too small?

$$e^{-\frac{i}{2(n+1)}} \cdot \operatorname{vol}(B(0,R)) < \operatorname{vol}(B(0,r))$$

Hence,

$$i > 2(n+1) \ln \left(\frac{\operatorname{vol}(B(0,R))}{\operatorname{vol}(B(0,r))}\right)$$

$$= 2(n+1) \ln \left(n^n \delta^n \cdot \delta^{3n}\right)$$

$$= 8n(n+1) \ln(\delta) + 2(n+1)n \ln(n)$$

$$= \mathcal{O}(\operatorname{poly}(n, \langle a_{\max} \rangle))$$

$$\boxed{\operatorname{Imp}} = \frac{\operatorname{EADS} II}{\operatorname{O} \operatorname{Harald} \operatorname{Räcke}}$$
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### **Separation Oracle:**

Let  $K \subseteq \mathbb{R}^n$  be a convex set. A separation oracle for K is an algorithm A that gets as input a point  $x \in \mathbb{R}^n$  and either

- certifies that  $x \in K$ ,
- or finds a hyperplane separating *x* from *K*.

We will usually assume that A is a polynomial-time algorithm.

In order to find a point in *K* we need

- a guarantee that a ball of radius r is contained in K,
- an initial ball B(c, R) with radius R that contains K,
- a separation oracle for *K*.

The Ellipsoid algorithm requires  $O(\text{poly}(n) \cdot \log(R/r))$ iterations. Each iteration is polytime for a polynomial-time Separation oracle.