#### How do we detect whether the LP is unbounded?

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

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

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#### Issues/Ouestions:

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

### **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  $\it K$  from  $\it z$  (e.g. a violated constraint in the LP).

► 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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- ▶ How do you measure progress? By how much does the

#### **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.



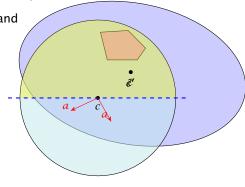
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## **How to Compute the New Ellipsoid**

- Use  $f^{-1}$  (recall that f = Lx + t is the transformation function for the Ellipsoid) 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  $\hat{c}'$  and the new matrix  $\hat{Q}'$  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*.



#### **Definition 5**

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

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

$$f(B(0,1)) = \{ f(x) \mid x \in B(0,1) \}$$

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

$$= \{ y \in \mathbb{R}^n \mid (y-t)^t L^{-1}^t L^{-1}(y-t) \le 1 \}$$

$$= \{ y \in \mathbb{R}^n \mid (y-t)^t Q^{-1}(y-t) \le 1 \}$$

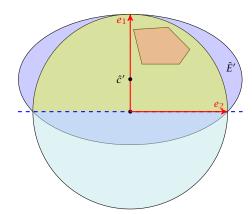
where  $Q = LL^t$  is an invertible matrix.



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### **The Easy Case**



- ▶ The new center lies on axis  $x_1$ . Hence,  $\hat{c}' = te_1$  for t > 0.
- ► The vectors  $e_1, e_2,...$  have to fulfill the ellipsoid constraint with equality. Hence  $(e_i \hat{c}')^t \hat{Q}'^{-1} (e_i \hat{c}') = 1$ .

## The Easy Case

- ▶ The obtain the matrix  $\hat{O}'^{-1}$  for our ellipsoid  $\hat{E}'$  note that  $\hat{E}'$ is axis-parallel.
- $\blacktriangleright$  Let a denote the radius along the  $x_1$ -axis and let b denote the (common) radius for the other axes.
- ▶ The matrix

$$\hat{L}' = \left(\begin{array}{cccc} a & 0 & \dots & 0 \\ 0 & b & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & b \end{array}\right)$$

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

 $(e_1 - \hat{c}')^t \hat{Q}'^{-1} (e_1 - \hat{c}') = 1$  gives

$$\begin{pmatrix} 1-t \\ 0 \\ \vdots \\ 0 \end{pmatrix}^{t} \cdot \begin{pmatrix} \frac{1}{a^{2}} & 0 & \dots & 0 \\ 0 & \frac{1}{b^{2}} & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \frac{1}{b^{2}} \end{pmatrix} \cdot \begin{pmatrix} 1-t \\ 0 \\ \vdots \\ 0 \end{pmatrix} = 1$$

► This gives  $(1 - t)^2 = a^2$ .

## The Easy Case

As  $\hat{O}' = \hat{L}'\hat{L}'^t$  the matrix  $\hat{O}'^{-1}$  is of the form

$$\hat{Q}'^{-1} = \begin{pmatrix} \frac{1}{a^2} & 0 & \dots & 0 \\ 0 & \frac{1}{b^2} & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & \frac{1}{b^2} \end{pmatrix}$$

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## 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

$$\frac{1}{h^2} = 1 - \frac{t^2}{a^2} = 1 - \frac{t^2}{(1-t)^2} = \frac{1-2t}{(1-t)^2}$$

# **Summary**

So far we have

$$a = 1 - t$$
 and  $b = \frac{1 - t}{\sqrt{1 - 2t}}$ 

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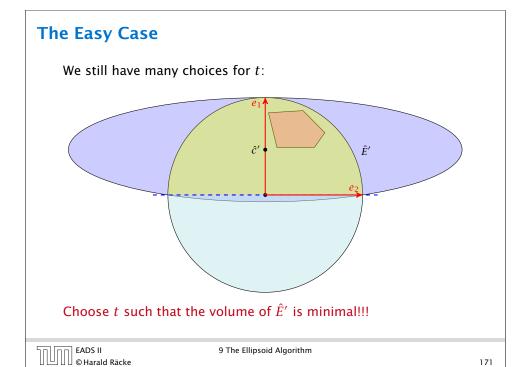
# **The Easy Case**

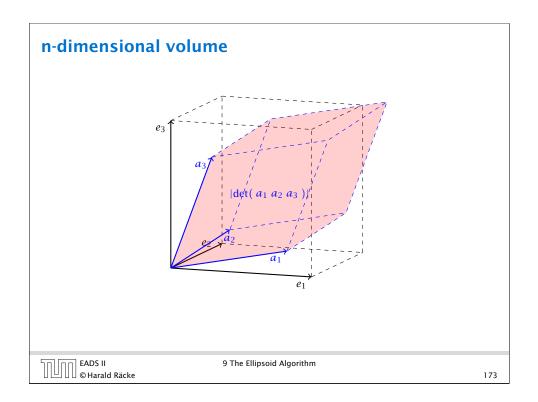
We want to choose t such that the volume of  $\hat{E}'$  is minimal.

#### Lemma 6

Let L be an affine transformation and  $K \subseteq \mathbb{R}^n$ . Then

$$vol(L(K)) = |det(L)| \cdot vol(K)$$
.





### **The Easy Case**

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

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

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

$$\begin{split} \operatorname{vol}(\hat{E}') &= \operatorname{vol}(B(0,1)) \cdot |\det(\hat{L}')| \\ &= \operatorname{vol}(B(0,1)) \cdot ab^{n-1} \\ &= \operatorname{vol}(B(0,1)) \cdot (1-t) \cdot \left(\frac{1-t}{\sqrt{1-2t}}\right)^{n-1} \\ &= \operatorname{vol}(B(0,1)) \cdot \frac{(1-t)^n}{(\sqrt{1-2t})^{n-1}} \end{split}$$

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# **The Easy Case**

$$\frac{\operatorname{d}\operatorname{vol}(\hat{E}')}{\operatorname{d}t} = \frac{\operatorname{d}}{\operatorname{d}t} \left( \frac{(1-t)^n}{(\sqrt{1-2t})^{n-1}} \right)$$

$$= \frac{1}{N^2} \cdot \left( (-1) \cdot n(1-t)^{n-1} \cdot (\sqrt{1-2t})^{n-1} \right)$$

$$= \frac{1}{N^2} \cdot \left( (-1) \cdot n(1-t)^{n-1} \cdot (\sqrt{1-2t})^{n-2} \cdot \frac{1}{2\sqrt{1-2t}} \cdot (\sqrt{1-2t})^{n-2} \cdot \frac{1}{2\sqrt{1-2t}} \cdot (\sqrt{1-2t})^{n-2} \cdot \frac{1}{2\sqrt{1-2t}} \cdot \left( (n-1)(1-t) - n(1-2t) \right)$$

$$= \frac{1}{N^2} \cdot (\sqrt{1-2t})^{n-3} \cdot (1-t)^{n-1} \cdot \left( (n+1)t-1 \right)$$

### **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:

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

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

$$\le e^{-2\frac{1}{n+1}} \cdot e^{\frac{1}{n+1}}$$

$$= e^{-\frac{1}{n+1}}$$

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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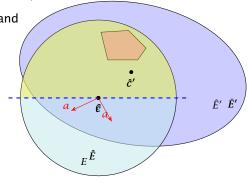
#### Our progress is the same:

$$e^{-\frac{1}{2(n+1)}} \ge \frac{\operatorname{vol}(\hat{E}')}{\operatorname{vol}(B(0,1))} = \frac{\operatorname{vol}(\hat{E}')}{\operatorname{vol}(\hat{E})} = \frac{\operatorname{vol}(R(\hat{E}'))}{\operatorname{vol}(R(\hat{E}))}$$
$$= \frac{\operatorname{vol}(\bar{E}')}{\operatorname{vol}(\bar{E})} = \frac{\operatorname{vol}(f(\bar{E}'))}{\operatorname{vol}(f(\bar{E}))} = \frac{\operatorname{vol}(E')}{\operatorname{vol}(E)}$$

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  $\hat{c}'$  and the new matrix  $\hat{Q}'$  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  $\bar{a} = L^t a$ .

### The Ellipsoid Algorithm

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

$$R^{-1}\left(\frac{L^t a}{\|L^t a\|}\right) = -e_1 \quad \Rightarrow \quad -\frac{L^t a}{\|L^t a\|} = R \cdot e_1$$

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

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

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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# 9 The Ellipsoid Algorithm

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

### 9 The Ellipsoid Algorithm

Hence,

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

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$$\begin{split} E' &= L(\bar{E}') \\ &= \{ L(x) \mid x^t \bar{Q}'^{-1} x \le 1 \} \\ &= \{ y \mid (L^{-1} y)^t \bar{Q}'^{-1} L^{-1} y \le 1 \} \\ &= \{ y \mid y^t (L^t)^{-1} \bar{Q}'^{-1} L^{-1} y \le 1 \} \\ &= \{ y \mid y^t (\underline{L} \bar{Q}' L^t)^{-1} y \le 1 \} \end{split}$$

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# 9 The Ellipsoid Algorithm

Hence,

$$\begin{aligned} Q' &= L\bar{Q}'L^t \\ &= L \cdot \frac{n^2}{n^2 - 1} \Big( I - \frac{2}{n+1} \frac{L^t a a^t L}{a^t Q a} \Big) \cdot L^t \\ &= \frac{n^2}{n^2 - 1} \Big( Q - \frac{2}{n+1} \frac{Q a a^t Q}{a^t Q a} \Big) \end{aligned}$$

# **Incomplete Algorithm**

#### Algorithm 1 ellipsoid-algorithm

1: **input:** point  $c \in \mathbb{R}^n$ , convex set  $K \subseteq \mathbb{R}^n$ 

2: **output:** point  $x \in K$  or "K is empty"

3: *Q* ← ???

4: repeat

5: **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:  $Q \leftarrow \frac{n^2}{n^2 - 1} \left( Q - \frac{2}{n+1} \frac{Qaa^t Q}{a^t Qa} \right)$ 

10: **endif** 

11: until ???

12: **return** "K is empty"

#### Repeat: Size of basic solutions

#### Lemma 7

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

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

Note that here we have  $P = \{x \mid Ax \leq b\}$ . 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  ${\it P}$  is bounded.

In this case 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  $R^n B(0,1) \le (n\delta)^n B(0,1)$ .

#### Repeat: Size of basic solutions

#### Proof:

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

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

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

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

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

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

#### Lemma 8

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

⇔: obvious!

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

 $ar{P}_{\lambda}$  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

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

⇒:

Consider the polytops

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

and

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

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

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

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  $r^n \text{vol}(B(0,1)=\frac{1}{\delta^3 n} \text{vol}(B(0,1))$ .

#### **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

$$(A(x + \vec{\ell}))_i = (Ax)_i + (A\vec{\ell})_i \le b_i + A_i \vec{\ell}$$

$$\le b_i + ||A_i|| \cdot ||\vec{\ell}|| \le b_i + \sqrt{n} \cdot 2^{\langle a_{\text{max}} \rangle} \cdot r$$

$$\le b_i + \frac{\sqrt{n} \cdot 2^{\langle a_{\text{max}} \rangle}}{\delta^3} \le b_i + \frac{1}{\delta^2 + 1} \le b_i + \frac{1}{\lambda}$$

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
- 2: with  $K \subseteq B(0,R)$ , and  $B(x,r) \subseteq K$  for some x
- 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:  $c \leftarrow 0$
- 6: repeat
- 7: if  $c \in K$  then return c
- 8: else
- 9: choose a violated hyperplane *a*
- 10:  $c \leftarrow c \frac{1}{n+1} \frac{Qa}{\sqrt{a^t Qa}}$
- 11:  $Q \leftarrow \frac{n^2}{n^2 1} \left( Q \frac{2}{n+1} \frac{Qaa^t Q}{a^t Qa} \right)$
- 12: endif
- 13: **until**  $det(Q) \le r^{2n}$  // i.e.,  $det(L) \le r^n$
- 14: 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,

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

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

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

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

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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

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

We will usually assume that  $\boldsymbol{A}$  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,
- $\blacktriangleright$  an initial ball B(c,R) with radius R that contains K,
- ▶ a separation oracle for *K*.

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

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