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?
- ▶ 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

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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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.

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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.



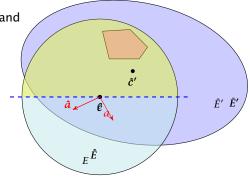
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How to Compute the New Ellipsoid

- 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.
- ▶ 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 $x = L^{-1}(f(x) - t)$.

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

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

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

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

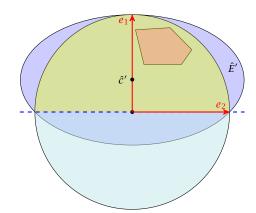
where $Q = LL^t$ is an invertible matrix.



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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 $\hat{Q'}^{-1}$ for our ellipsoid \hat{E}' note that \hat{E}' is axis-parallel.
- 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 $\hat{Q}' = \hat{L}' \hat{L}'^t$ the matrix \hat{Q}'^{-1} is of the form

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

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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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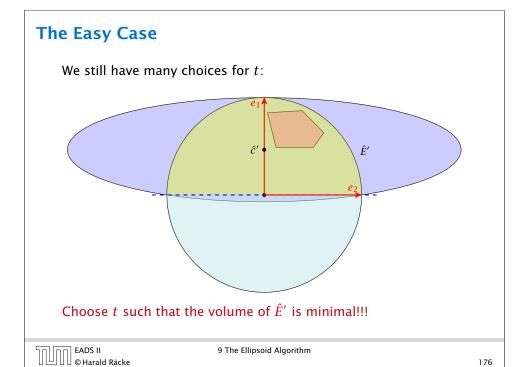
The Easy Case

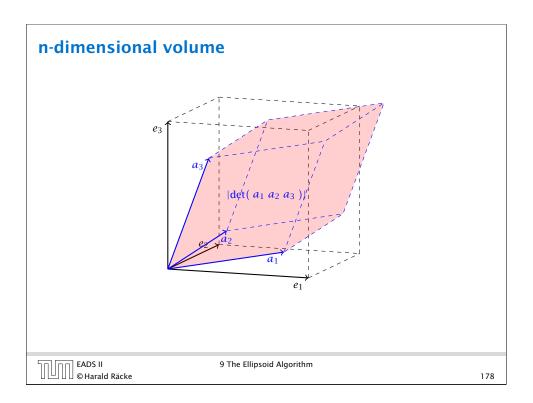
We want to choose t such that the volume of \hat{E}' is minimal.

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 $\hat{O}' = \hat{L}' \hat{L}'^t$.

We have

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

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

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The Easy Case

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

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

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

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

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

The Easy Case

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

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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:

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

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

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

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

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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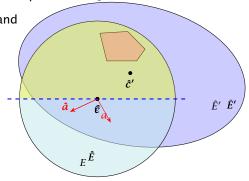
Our progress is the same:

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

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

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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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Hence,

$$Q' = L\bar{Q}'L^t$$

$$= L \cdot \frac{n^2}{n^2 - 1} \left(I - \frac{2}{n+1} \frac{L^t a a^t L}{a^t Q a} \right) \cdot L^t$$

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

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**

a: choose a violated hyperplane a

8: $c \leftarrow c - \frac{1}{n+1} \frac{Qa}{\sqrt{a^t Qa}}$

9: $Q \leftarrow \frac{n^2}{n^2 - 1} \left(Q - \frac{2}{n+1} \frac{Qaa^tQ}{a^tQa} \right)$

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, b. Then every entry x_j in a basic solution fulfills $|x_j| = \frac{D_j}{D}$ with $D_j, D \leq 2^{2n\langle a_{\max} \rangle + 2n\log_2 n}$.

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

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

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

$$\det(\bar{A}_B), \det(\bar{M}_j) \leq \|\vec{\ell}_{\max}\|^{2n}$$

$$\leq (\sqrt{2n} \cdot 2^{\langle a_{\max} \rangle})^{2n} \leq 2^{2n\langle a_{\max} \rangle + 2n\log_2 n},$$

where $\vec{\ell}_{\text{max}}$ 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 \bar{A} to \bar{A}_B do not increase the determinant. Only the at most 2n 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_{λ} is feasible if and only if P is feasible.

⇔: obvious!

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

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

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_{λ} feasible if and only if \bar{P}_{λ} feasible.

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

By Cramers rule we get

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

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

Proof:

If P_{λ} 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_{λ} 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 $K \subseteq B(c,R)$, and $B(x,r) \subseteq K$ for some x
- 3: **output:** point $x \in K$ or "K is empty"
- 4: $Q \leftarrow \text{diag}(R^2, ..., R^2) // \text{i.e., } L = \text{diag}(R, ..., R)$
- 5: repeat
- 6: if $c \in K$ then return c
- 7: else
- 8: choose a violated hyperplane *a*
- 9: $c \leftarrow c \frac{1}{n+1} \frac{Qa}{\sqrt{a^t Qa}}$
- 10: $Q \leftarrow \frac{n^2}{n^2 1} \left(Q \frac{2}{n+1} \frac{Qaa^t Q}{a^t Oa} \right)$
- 11: endi
- 12: **until** $det(Q) \le r^{2n}$ // i.e., $det(L) \le r^n$
- 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,

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

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

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

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

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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

- ightharpoonup 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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