

Ellipsoid Method

An ellipsoid with centre $c \in \mathbb{R}^n$ and a shape-size matrix $A \in \mathbb{S}_{++}^n$ is
 $E(A, c) := \{x \in \mathbb{R}^n; \langle A^{-1}(x-c), (x-c) \rangle \leq 1\}$.

$E(A, c)$ is the image of Euclidean Unit Ball $B_n(0, 1)$.

$$E(A, c) = c + A^{1/2} B_n(0, 1).$$

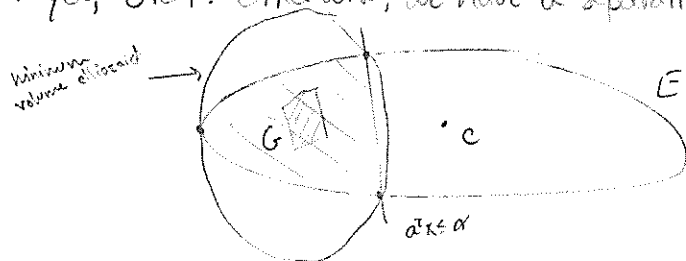
$$\text{vol}(E(A, c)) = \sqrt{\det(A)} \text{vol}(B_n(0, 1))$$

Ellipsoids are simple convex sets that are easy to deal with in mathematical analysis and in algorithms. Also, they are rich enough to approximate every convex set relatively well:

Theorem 3.1: For every compact convex set G in \mathbb{R}^d with non-empty interior, \exists a unique minimal volume ellipsoid which contains G . Moreover, shrinking the ellipsoid around its centre by a factor of d yields an ellipsoid contained in G .
 (Pod hints in textbook.) (Löwner-John ellipsoid)

Sketch of finding a point of a convex set G via Ellipsoid Method.

1. Find a possibly very large Ellipsoid E which contains G .
2. Ask a separation oracle for G : is the center of the Ellipsoid E in G ?
3. If yes, STOP. Otherwise, we have a separating hyperplane:



4. Construct the minimum volume Ellipsoid which contains $E \cap \{x; \alpha x \leq \alpha\}$.
 GO TO 2 with this new Ellipsoid in place of E .

To summarize, we will see a way of applying Ellipsoid Method for solutions of

$$\inf_{x \in G} c^T x$$

where $G \subset \mathbb{R}^d$ is convex and bounded with $\text{int}(G) \neq \emptyset$.

(assume) \rightarrow There are $r, R > 0$ such that for some $\tilde{x} \in G$,

$$B_d(\tilde{x}, r) \subseteq G \subseteq B_d(\tilde{x}, R)$$

Given $\epsilon > 0$, define

$$\mu_0 := \epsilon + \sup\{c^T x; x \in G\} - \inf\{c^T x; x \in G\}.$$

Then Ellipsoid Method can be used to generate a feasible solution $\bar{x} \in G$ such that

$$c^T \bar{x} \leq \inf\{c^T x; x \in G\} + \epsilon$$

in

$$O\left(d^2 \left(\ln\left(\frac{R}{r}\right) + \ln\left(\frac{\mu_0}{\epsilon}\right)\right)\right)$$

iterations, where each iteration makes a call to the separation oracle for G and does some $O(d^2)$ computations.

Let's apply the last result to SDP. Suppose (P) and (D) have Slater points \bar{X} and (\bar{s}, \bar{S}) respectively. Consider

$$\inf \langle C, X \rangle$$

$$(P) \quad A(X) = b$$

$$\langle \bar{S}, X \rangle \leq 2 \langle \bar{S}, \bar{X} \rangle$$

$$X \in \mathcal{S}_+^n$$

Theorem 3.9:

(a) (P) & (P̃) attain their optimal values.

(b) The optimal solution sets of (P) & (P̃) are the same.

(c) Let G denote the feasible region of (P̃). Then G is convex and

$$B_{\dim(G)}(\bar{X}, \lambda_n(\bar{X})) \subseteq G \subseteq B_{\dim(G)}\left(0, \frac{2\langle \bar{X}, \bar{S} \rangle}{\lambda_n(\bar{S})}\right).$$