

15.081J/6.251J Introduction to Mathematical  
Programming

Lecture 21: Primal Barrier  
Interior Point Algorithm

# 1 Outline

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1. Barrier Methods
2. The Central Path
3. Approximating the Central Path
4. The Primal Barrier Algorithm
5. Correctness and Complexity

# 2 Barrier methods

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$$\begin{aligned} \min \quad & f(\mathbf{x}) \\ \text{s.t.} \quad & g_j(\mathbf{x}) \leq 0, \quad j = 1, \dots, p \\ & h_i(\mathbf{x}) = 0, \quad i = 1, \dots, m \end{aligned}$$

$$\begin{aligned} S = \{ \mathbf{x} \mid & g_j(\mathbf{x}) < 0, \quad j = 1, \dots, p, \\ & h_i(\mathbf{x}) = 0, \quad i = 1, \dots, m \} \end{aligned}$$

## 2.1 Strategy

SLIDE 3

- A barrier function  $G(\mathbf{x})$  is a continuous function with the property that it approaches  $\infty$  as one of  $g_j(\mathbf{x})$  approaches 0 from negative values.
- Examples:

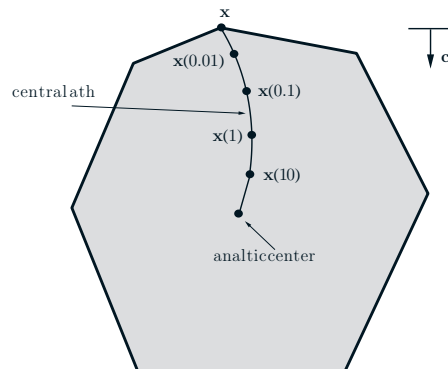
$$G(\mathbf{x}) = - \sum_{j=1}^p \log(-g_j(\mathbf{x})), \quad G(\mathbf{x}) = - \sum_{j=1}^p \frac{1}{g_j(\mathbf{x})}$$

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- Consider a sequence of  $\mu^k$ :  $0 < \mu^{k+1} < \mu^k$  and  $\mu^k \rightarrow 0$ .
- Consider the problem

$$\mathbf{x}^k = \operatorname{argmin}_{\mathbf{x} \in S} \{ f(\mathbf{x}) + \mu^k G(\mathbf{x}) \}$$

- Theorem Every limit point  $\mathbf{x}^k$  generated by a barrier method is a global minimum of the original constrained problem.



## 2.2 Primal path-following IPMs for LO

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$$\begin{array}{ll}
 (P) \min & \mathbf{c}'\mathbf{x} \\
 \text{s.t.} & \mathbf{Ax} = \mathbf{b} \\
 & \mathbf{x} \geq \mathbf{0}
 \end{array}
 \qquad
 \begin{array}{ll}
 (D) \max & \mathbf{b}'\mathbf{p} \\
 \text{s.t.} & \mathbf{A}'\mathbf{p} + \mathbf{s} = \mathbf{c} \\
 & \mathbf{s} \geq \mathbf{0}
 \end{array}$$

Barrier problem:

$$\begin{array}{ll}
 \min & B_\mu(\mathbf{x}) = \mathbf{c}'\mathbf{x} - \mu \sum_{j=1}^n \log x_j \\
 \text{s.t.} & \mathbf{Ax} = \mathbf{b}
 \end{array}$$

Minimizer:  $\mathbf{x}(\mu)$

## 3 Central Path

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- As  $\mu$  varies, minimizers  $\mathbf{x}(\mu)$  form the central path
- $\lim_{\mu \rightarrow 0} \mathbf{x}(\mu)$  exists and is an optimal solution  $\mathbf{x}^*$  to the initial LP
- For  $\mu = \infty$ ,  $\mathbf{x}(\infty)$  is called the *analytic center*

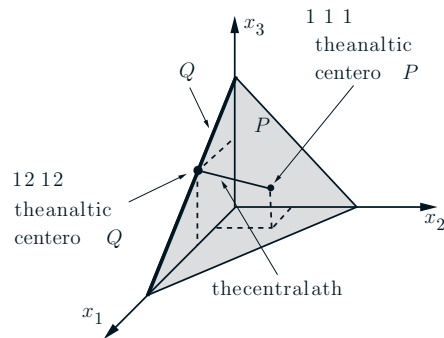
$$\begin{array}{ll}
 \min & - \sum_{j=1}^n \log x_j \\
 \text{s.t.} & \mathbf{Ax} = \mathbf{b}
 \end{array}$$

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### 3.1 Example

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$$\begin{array}{ll}
 \min & x_2 \\
 \text{s.t.} & x_1 + x_2 + x_3 = 1 \\
 & x_1, x_2, x_3 \geq 0
 \end{array}$$



- $Q = \{\mathbf{x} \mid \mathbf{x} = (x_1, 0, x_3), x_1 + x_3 = 1, \mathbf{x} \geq \mathbf{0}\}$ , set of optimal solutions to original LP
- The analytic center of  $Q$  is  $(1/2, 0, 1/2)$

$$\begin{aligned} \min \quad & x_2 - \mu \log x_1 - \mu \log x_2 - \mu \log x_3 \\ \text{s.t.} \quad & x_1 + x_2 + x_3 = 1 \end{aligned}$$

$$\min \quad x_2 - \mu \log x_1 - \mu \log x_2 - \mu \log(1 - x_1 - x_2).$$

$$\begin{aligned} x_1(\mu) &= \frac{1 - x_2(\mu)}{2} \\ x_2(\mu) &= \frac{1 + 3\mu - \sqrt{1 + 9\mu^2 + 2\mu}}{2} \\ x_3(\mu) &= \frac{1 - x_2(\mu)}{2} \end{aligned}$$

The analytic center:  $(1/3, 1/3, 1/3)$

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### 3.2 Solution of Central Path

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- Barrier problem for dual:

$$\begin{aligned} \max \quad & \mathbf{p}'\mathbf{b} + \mu \sum_{j=1}^n \log s_j \\ \text{s.t.} \quad & \mathbf{p}'\mathbf{A} + \mathbf{s}' = \mathbf{c}' \end{aligned}$$

- Solution (KKT):

$$\begin{aligned} \mathbf{A}\mathbf{x}(\mu) &= \mathbf{b} \\ \mathbf{x}(\mu) &\geq \mathbf{0} \\ \mathbf{A}'\mathbf{p}(\mu) + \mathbf{s}(\mu) &= \mathbf{c} \\ \mathbf{s}(\mu) &\geq \mathbf{0} \\ \mathbf{X}(\mu)\mathbf{S}(\mu)\mathbf{e} &= \mathbf{e}\mu \end{aligned}$$

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- Theorem: If  $\mathbf{x}^*$ ,  $\mathbf{p}^*$ , and  $\mathbf{s}^*$  satisfy optimality conditions, then they are optimal solutions to problems primal and dual barrier problems.
- Goal: Solve barrier problem

$$\begin{aligned} \min \quad & B_\mu(\mathbf{x}) = \mathbf{c}'\mathbf{x} - \mu \sum_{j=1}^n \log x_j \\ \text{s.t.} \quad & \mathbf{A}\mathbf{x} = \mathbf{b} \end{aligned}$$

## 4 Approximating the central path

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$$\begin{aligned} \frac{\partial B_\mu(\mathbf{x})}{\partial x_i} &= c_i - \frac{\mu}{x_i} \\ \frac{\partial^2 B_\mu(\mathbf{x})}{\partial x_i^2} &= \frac{\mu}{x_i^2} \\ \frac{\partial^2 B_\mu(\mathbf{x})}{\partial x_i \partial x_j} &= 0, \quad i \neq j \end{aligned}$$

Given a vector  $\mathbf{x} > \mathbf{0}$ :

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$$\begin{aligned} B_\mu(\mathbf{x} + \mathbf{d}) &\approx B_\mu(\mathbf{x}) + \sum_{i=1}^n \frac{\partial B_\mu(\mathbf{x})}{\partial x_i} d_i + \\ &\quad \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 B_\mu(\mathbf{x})}{\partial x_i \partial x_j} d_i d_j \\ &= B_\mu(\mathbf{x}) + (\mathbf{c}' - \mu \mathbf{e}' \mathbf{X}^{-1}) \mathbf{d} + \frac{1}{2} \mu \mathbf{d}' \mathbf{X}^{-2} \mathbf{d} \end{aligned}$$

$\mathbf{X} = \text{diag}(x_1, \dots, x_n)$

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Approximating problem:

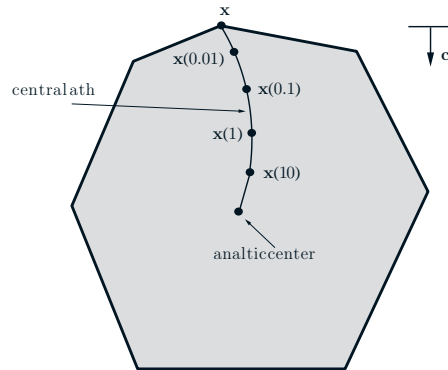
$$\begin{aligned} \min \quad & (\mathbf{c}' - \mu \mathbf{e}' \mathbf{X}^{-1}) \mathbf{d} + \frac{1}{2} \mu \mathbf{d}' \mathbf{X}^{-2} \mathbf{d} \\ \text{s.t.} \quad & \mathbf{A} \mathbf{d} = \mathbf{0} \end{aligned}$$

Solution (from Lagrange):

$$\begin{aligned} \mathbf{c} - \mu \mathbf{X}^{-1} \mathbf{e} + \mu \mathbf{X}^{-2} \mathbf{d} - \mathbf{A}' \mathbf{p} &= \mathbf{0} \\ \mathbf{A} \mathbf{d} &= \mathbf{0} \end{aligned}$$

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- System of  $m + n$  linear equations, with  $m + n$  unknowns ( $d_j$ ,  $j = 1, \dots, n$ , and  $p_i$ ,  $i = 1, \dots, m$ ).



- Solution:

$$d(\mu) = \left( I - X^2 A' (A X^2 A')^{-1} A \right) \left( x e - \frac{1}{\mu} X^2 c \right)$$

$$p(\mu) = (A X^2 A')^{-1} A (X^2 c - \mu x e)$$

#### 4.1 The Newton connection

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- $d(\mu)$  is the *Newton direction*; process of calculating this direction is called a *Newton step*
- Starting with  $x$ , the new primal solution is  $x + d(\mu)$
- The corresponding dual solution becomes  $(p, s) = (p(\mu), c - A'p(\mu))$
- We then decrease  $\mu$  to  $\bar{\mu} = \alpha\mu$ ,  $0 < \alpha < 1$

#### 4.2 Geometric Interpretation

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- Take one Newton step so that  $x$  would be close to  $x(\mu)$
- Measure of closeness

$$\left\| \frac{1}{\mu} X S e - e \right\| \leq \beta,$$

$$0 < \beta < 1, \quad X = \text{diag}(x_1, \dots, x_n) \quad S = \text{diag}(s_1, \dots, s_n)$$

- As  $\mu \rightarrow 0$ , the complementarity slackness condition will be satisfied

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## 5 The Primal Barrier Algorithm

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Input

- (a)  $(\mathbf{A}, \mathbf{b}, \mathbf{c})$ ;  $\mathbf{A}$  has full row rank;
- (b)  $\mathbf{x}^0 > \mathbf{0}$ ,  $\mathbf{s}^0 > \mathbf{0}$ ,  $\mathbf{p}^0$ ;
- (c) optimality tolerance  $\epsilon > 0$ ;
- (d)  $\mu^0$ , and  $\alpha$ , where  $0 < \alpha < 1$ .

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1. (Initialization) Start with some primal and dual feasible  $\mathbf{x}^0 > \mathbf{0}$ ,  $\mathbf{s}^0 > \mathbf{0}$ ,  $\mathbf{p}^0$ , and set  $k = 0$ .
2. (Optimality test) If  $(\mathbf{s}^k)' \mathbf{x}^k < \epsilon$  stop; else go to Step 3.
3. Let

$$\begin{aligned}\mathbf{X}_k &= \text{diag}(x_1^k, \dots, x_n^k), \\ \mu^{k+1} &= \alpha \mu^k\end{aligned}$$

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4. (Computation of directions) Solve the linear system

$$\begin{aligned}\mu^{k+1} \mathbf{X}_k^{-2} \mathbf{d} - \mathbf{A}' \mathbf{p} &= \mu^{k+1} \mathbf{X}_k^{-1} \mathbf{e} - \mathbf{c} \\ \mathbf{A} \mathbf{d} &= \mathbf{0}\end{aligned}$$

5. (Update of solutions) Let

$$\begin{aligned}\mathbf{x}^{k+1} &= \mathbf{x}^k + \mathbf{d}, \\ \mathbf{p}^{k+1} &= \mathbf{p}, \\ \mathbf{s}^{k+1} &= \mathbf{c} - \mathbf{A}' \mathbf{p}.\end{aligned}$$

6. Let  $k := k + 1$  and go to Step 2.

## 6 Correctness

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Theorem Given  $\alpha = 1 - \frac{\sqrt{\beta} - \beta}{\sqrt{\beta} + \sqrt{n}}$ ,  $\beta < 1$ ,  $(\mathbf{x}^0, \mathbf{s}^0, \mathbf{p}^0)$ ,  $(\mathbf{x}^0 > \mathbf{0}, \mathbf{s}^0 > \mathbf{0})$ :

$$\left\| \frac{1}{\mu^0} \mathbf{X}_0 \mathbf{S}_0 \mathbf{e} - \mathbf{e} \right\| \leq \beta.$$

Then, after

$$K = \left\lceil \frac{\sqrt{\beta} + \sqrt{n}}{\sqrt{\beta} - \beta} \log \frac{(\mathbf{s}^0)' \mathbf{x}^0 (1 + \beta)}{\epsilon (1 - \beta)} \right\rceil$$

iterations,  $(\mathbf{x}^K, \mathbf{s}^K, \mathbf{p}^K)$  is found:

$$(\mathbf{s}^K)' \mathbf{x}^K \leq \epsilon.$$

## 6.1 Proof

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- Claim (by induction):  $\left| \frac{1}{\mu^k} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right| \leq \beta$
- For  $k = 0$  we have assumed it
- Assume it holds for  $k$ ;

$$\begin{aligned}
 \left| \frac{1}{\mu^{k+1}} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right| &= \left| \frac{1}{\alpha \mu^k} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right| \\
 &= \left| \frac{1}{\alpha} \left( \frac{1}{\mu^k} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right) + \frac{1-\alpha}{\alpha} \mathbf{e} \right| \\
 &\leq \frac{1}{\alpha} \left| \frac{1}{\mu^k} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right| + \frac{1-\alpha}{\alpha} \|\mathbf{e}\| \\
 &\leq \frac{\beta}{\alpha} + \frac{1-\alpha}{\alpha} \sqrt{n} \\
 &= \sqrt{\beta}
 \end{aligned}$$

- We next show that  $\|\mathbf{X}_k^{-1} \mathbf{d}\| \leq \sqrt{\beta} < 1$ , where  $\mathbf{d} = \mathbf{x}^{k+1} - \mathbf{x}^k$ .
- $\mathbf{d}$  solves

$$\begin{aligned}
 \mu^{k+1} \mathbf{X}_k^{-2} \mathbf{d} - \mathbf{A}' \mathbf{p} &= \mu^{k+1} \mathbf{X}_k^{-1} \mathbf{e} - \mathbf{c}, \\
 \mathbf{A} \mathbf{d} &= \mathbf{0}
 \end{aligned}$$

- By left-multiplying the first equation by  $\mathbf{d}'$

$$\mu^{k+1} \mathbf{d}' \mathbf{X}_k^{-2} \mathbf{d} = \mathbf{d}' \left( \mu^{k+1} \mathbf{X}_k^{-1} \mathbf{e} - \mathbf{c} \right)$$

$$\begin{aligned}
 \|\mathbf{X}_k^{-1} \mathbf{d}\|^2 &= \mathbf{d}' \mathbf{X}_k^{-2} \mathbf{d} \\
 &= \left( \mathbf{X}_k^{-1} \mathbf{e} - \frac{1}{\mu^{k+1}} \mathbf{c} \right)' \mathbf{d} \\
 &= \left( \mathbf{X}_k^{-1} \mathbf{e} - \frac{1}{\mu^{k+1}} (\mathbf{s}^k + \mathbf{A}' \mathbf{p}^k) \right)' \mathbf{d} \\
 &= \left( \mathbf{X}_k^{-1} \mathbf{e} - \frac{1}{\mu^{k+1}} \mathbf{s}^k \right)' \mathbf{d} \\
 &= - \left( \frac{1}{\mu^{k+1}} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right)' \mathbf{X}_k^{-1} \mathbf{d} \\
 &\leq \left| \frac{1}{\mu^{k+1}} \mathbf{X}_k \mathbf{S}_k \mathbf{e} - \mathbf{e} \right| \|\mathbf{X}_k^{-1} \mathbf{d}\| \\
 &\leq \sqrt{\beta} \|\mathbf{X}_k^{-1} \mathbf{d}\|
 \end{aligned}$$

hence,  $\|\mathbf{X}_k^{-1} \mathbf{d}\| \leq \sqrt{\beta} < 1$ .



- We next show that  $\mathbf{x}^{k+1}$  and  $(\mathbf{p}^{k+1}, \mathbf{s}^{k+1})$  are primal and dual feasible. Since  $\mathbf{A}\mathbf{d} = \mathbf{0}$ , we have

$$\begin{aligned} \mathbf{A}\mathbf{x}^{k+1} &= \mathbf{b} \\ \mathbf{x}^{k+1} &= \mathbf{x}^k + \mathbf{d} = \mathbf{X}_k(\mathbf{e} + \mathbf{X}_k^{-1}\mathbf{d}) > \mathbf{0}, \end{aligned}$$

because  $\|\mathbf{X}_k^{-1}\mathbf{d}\| < 1$

$$\mathbf{A}'\mathbf{p}^{k+1} + \mathbf{s}^{k+1} = \mathbf{c},$$

by construction and

$$\mathbf{s}^{k+1} = \mathbf{c} - \mathbf{A}'\mathbf{p}^{k+1} = \mu^{k+1}\mathbf{X}_k^{-1}(\mathbf{e} - \mathbf{X}_k^{-1}\mathbf{d}) > \mathbf{0},$$

because  $\|\mathbf{X}_k^{-1}\mathbf{d}\| < 1$

•

$$\begin{aligned} x_j^{k+1} &= x_j^k \left(1 + \frac{d_j}{x_j^k}\right), \\ s_j^{k+1} &= \frac{\mu^{k+1}}{x_j^k} \left(1 - \frac{d_j}{x_j^k}\right). \end{aligned}$$

Therefore,

$$\begin{aligned} \frac{1}{\mu^{k+1}}x_j^{k+1}s_j^{k+1} - 1 &= \frac{1}{\mu^{k+1}}x_j^k \left(1 + \frac{d_j}{x_j^k}\right) \frac{\mu^{k+1}}{x_j^k} \left(1 - \frac{d_j}{x_j^k}\right) - 1 \\ &= -\left(\frac{d_j}{x_j^k}\right)^2. \end{aligned}$$

- $\mathbf{D} = \text{diag}(d_1, \dots, d_n)$ ,  $\|\mathbf{u}\|_1 = \sum_i |u_i|$ . Note that  $\|\mathbf{u}\| \leq \|\mathbf{u}\|_1$

$$\begin{aligned} \left| \frac{1}{\mu^{k+1}}\mathbf{X}_{k+1}\mathbf{S}_{k+1}\mathbf{e} - \mathbf{e} \right| &= \|\mathbf{X}_k^{-2}\mathbf{D}^2\mathbf{e}\| \\ &\leq \|\mathbf{X}_k^{-2}\mathbf{D}^2\mathbf{e}\|_1 \\ &= \mathbf{e}'\mathbf{X}_k^{-2}\mathbf{D}^2\mathbf{e} \\ &= \mathbf{e}'\mathbf{D}\mathbf{X}_k^{-2}\mathbf{D}\mathbf{e} \\ &= \mathbf{d}'\mathbf{X}_k^{-2}\mathbf{d} \\ &= \|\mathbf{X}_k^{-1}\mathbf{d}\|^2 \\ &\leq (\sqrt{\beta})^2 \\ &= \beta, \end{aligned}$$

and hence the induction is complete.

- Since at every iteration

$$\begin{aligned} \left| \frac{1}{\mu^k}\mathbf{X}_k\mathbf{S}_k\mathbf{e} - \mathbf{e} \right| &\leq \beta \\ -\beta &\leq \frac{1}{\mu^k}x_j^k s_j^k - 1 \leq \beta \\ n\mu^k(1 - \beta) &\leq (\mathbf{s}^k)'\mathbf{x}^k \leq n\mu^k(1 + \beta) \end{aligned}$$

- 

$$\mu^k = \alpha^k \mu^0 = \left(1 - \frac{\sqrt{\beta} - \beta}{\sqrt{\beta} + \sqrt{n}}\right)^k \mu^0 \leq e^{-k \frac{\sqrt{\beta} - \beta}{\sqrt{\beta} + \sqrt{n}}} \mu^0$$

- After

$$\left\lceil \frac{\sqrt{\beta} + \sqrt{n}}{\sqrt{\beta} - \beta} \log \frac{\mu^0 n(1 + \beta)}{\epsilon} \right\rceil \leq \left\lceil \frac{\sqrt{\beta} + \sqrt{n}}{\sqrt{\beta} - \beta} \log \frac{(\mathbf{s}^0)' \mathbf{x}^0 (1 + \beta)}{\epsilon(1 - \beta)} \right\rceil = K$$

iterations, the primal barrier algorithm finds primal and dual solutions  $\mathbf{x}^K$ ,  $(\mathbf{p}^K, \mathbf{s}^K)$ , that have duality gap  $(\mathbf{s}^K)' \mathbf{x}^K$  less than or equal to  $\epsilon$

## 7 Complexity

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- Work per iteration involves solving a linear system with  $m + n$  equations in  $m + n$  unknowns. Given that  $m \leq n$ , the work per iteration is  $O(n^3)$ .
- $\epsilon_0 = (\mathbf{s}^0)' \mathbf{x}^0$ : initial duality gap. Algorithm needs

$$O\left(\sqrt{n} \log \frac{\epsilon_0}{\epsilon}\right)$$

iterations to reduce the duality gap from  $\epsilon_0$  to  $\epsilon$ , with  $O(n^3)$  arithmetic operations per iteration.

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