

# Problem

$Q \subset \mathbb{R}^n$  - open convex set, bounded

$$\min \{ \langle c, x \rangle, x \in \bar{Q} \}$$

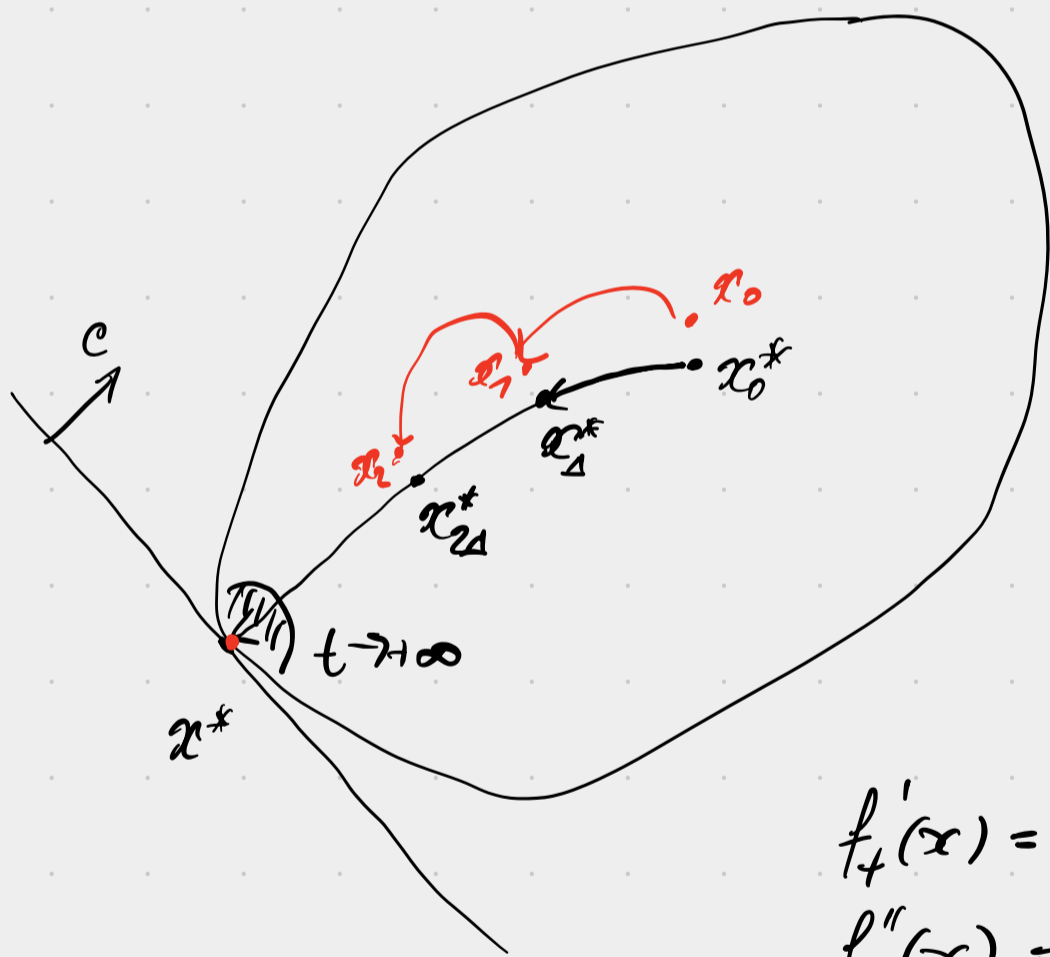
Introduce,  $t \geq 0$

$$f_t(x) = t \langle c, x \rangle + F(x), \quad F - \text{a "barrier" for } Q$$

$\text{dom } F = Q$

Define  $x_t^* = \underset{x \in Q}{\text{argmin}} f_t(x) = \underset{x \in \text{dom } F}{\text{argmin}} f_t(x)$  - central path

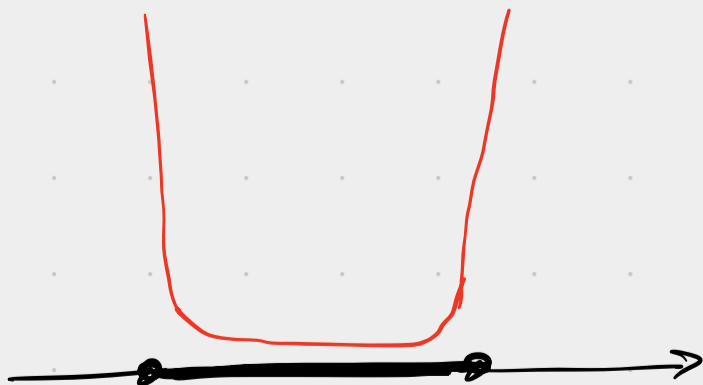
$x_0^* = \underset{x \in Q}{\text{argmin}} F(x)$  - an "analytic center" of  $Q$ .



$$x_t^* \in Q$$

$\Delta$  - small, not tiny

$$f_t'(x) = tc + F'(x)$$
$$f_t''(x) = \underline{\underline{F''(x)}}$$



## Self-Concordant Barriers

Def

We say that  $F: Q \rightarrow \mathbb{R}$  is a self-concordant barrier for set  $Q$  with parameter  $\theta > 0$ :

1.  $Q$  is domain of  $F$ : For  $x_k \rightarrow \partial Q$ :  $F(x_k) \rightarrow +\infty$ .

2.  $F$  is a standard self-concordant function ( $M=2$ ):

$$F'''(x)[h, h, h] \leq 2 \|h\|_x^3 \equiv 2 (F''(x)[h, h])^{3/2}$$

$\forall x \in Q$   
 $h \in \mathbb{R}^n$

3.  $F$  is Lipschitz w.r.t. local norm:

$$\|DF(x)\|_x^2 = \langle F'(x), F''(x)^{-1} F'(x) \rangle \leq \theta, \forall x \in Q.$$

Fact:  $x \in Q$ , the entire Dikin's ellipsoid belongs to  $Q$ :

$$E_x = \{y \in \mathbb{R}^n : \|y - x\|_x < 1\} \subset Q.$$

Main Result:

Theorem to find  $\langle c, x_n - x^* \rangle \leq \epsilon$

it's enough to perform

$$k = O(\sqrt{\theta} \log \frac{1}{\epsilon})$$

number of Newton's step.

## Equivalent Conditions

$$1. \langle F'(x), F''(x)^{-1} F'(x) \rangle \leq \theta \quad \forall x \in Q$$

$\Leftrightarrow$

$$2. \underbrace{2 \langle F'(x), u \rangle - \langle F''(x) u, u \rangle}_{\max u} \leq \theta \quad \forall u \in \mathbb{R}^n, \forall x \in Q$$

$F''(x) > 0 \Leftrightarrow Q$  does not contain lines

3. Homogenization:  $u = \tau h, \tau > 0, h \in \mathbb{R}^n$

$$\max_{\tau > 0} \left[ 2\tau \langle F'(x), h \rangle - \tau^2 \langle F''(x) h, h \rangle \right] = \frac{\langle F'(x), h \rangle^2}{\langle F''(x) h, h \rangle} \leq \theta$$

For any  $h \in \mathbb{R}^n$ :

$$\underline{\langle F'(x), h \rangle^2 \leq \theta \cdot \langle F''(x) h, h \rangle}$$

4.  $\forall x \in Q$ :

$$F''(x) \succeq \frac{1}{\theta} F'(x) F'(x)^T$$

"self-concordant notion of strong convexity"

Example 1.  $F(x) = -\log x$  is a self-concordant barrier,  $\theta = 1$

$$F'(x) = -\frac{1}{x}$$

$$F''(x) = \frac{1}{x^2}$$

$$\frac{[F'(x)]^2}{F''(x)} = 1 = \theta$$

Example 2.  $F(X) = -\log \det X$  is a self-concordant barrier

$$\theta = n$$

$$DF(X)[H] = -\text{tr}(X^{-1}H) = -\text{tr}(S),$$

$$S = X^{-1/2} H X^{-1/2} \text{ - symmetric}$$

$$D^2F(X)[H, H] = \text{tr}(X^{-1} H X^{-1} H) = \text{tr}(S^2)$$

$$(\text{tr}(S))^2 \stackrel{?}{\leq} \theta \text{tr}(S^2)$$

$$\left( \sum_{i=1}^n \lambda_i(S) \right)^2 \leq \theta \cdot \sum_{i=1}^n \lambda_i^2(S) \quad \theta = n$$

$$\|\cdot\|_1 \leq \sqrt{n} \cdot \|\cdot\|_2 \quad \square$$

## Basic Operations

Proposition  $F_1, \dots, F_m$  - self-concordant barriers for  $Q_1, \dots, Q_m \subseteq \mathbb{R}^n$

$\theta_1, \dots, \theta_m$

Then,

$$Q = \bigcap_{1 \leq i \leq m} Q_i$$

$$F(x) = \sum_{1 \leq i \leq m} F_i(x), \quad \theta = \sum_{i=1}^m \theta_i.$$

Proof

$$\begin{aligned} & \max_{u \in \mathbb{R}^n} \left[ 2 \langle F'(x), u \rangle - \langle F''(x)u, u \rangle \right] = \\ & = \max_{u \in \mathbb{R}^n} \left[ \sum_{i=1}^m \left( 2 \langle F_i'(x), u \rangle - \langle F_i''(x)u, u \rangle \right) \right] \\ & \leq \sum_{i=1}^m \max_{u \in \mathbb{R}^n} [-1] \leq \sum_{i=1}^m \theta_i = \theta. \quad \square \end{aligned}$$

Example  $Q = \{ x \in \mathbb{R}^n \mid x_i > 0, 1 \leq i \leq n \} = \mathbb{R}_{>0}^n$ .

$$F(x) = - \sum_{i=1}^n \log x_i$$

$$\boxed{\theta = n}$$

## Affine Invariance $A(x) = Ax + b$

let  $F: Q \rightarrow \mathbb{R}$  - self-concordant barrier for  $Q, \theta_F$ .

$\Phi(x) = F(A(x))$  - self-concordant barrier for

$$\Omega = \{x : A(x) \in Q\} = \{x \in A^{-1}(Q)\}.$$

$$\boxed{\theta_\Phi = \theta_F}$$

## Example Linear Programming

$$Q = \{ \langle a_1, x \rangle \leq b_1, \dots, \langle a_m, x \rangle \leq b_m \}$$

$$F(x) = - \sum_{i=1}^m \log(b_i - \langle a_i, x \rangle)$$

$$\boxed{\theta = m}$$

$$\min_{\substack{f(x) \\ \underline{f(x) \leq 0}}} \} \rightarrow \min_x t f(x) - \log(-f(x))$$

"Barrier approach"

CVX, Mosek.

$$\min_x t f(x) + [f(x)]_+^2$$

"Penalty Function"

## Key Property

Lemma  $x, y \in Q$

$$\langle F'(x), y-x \rangle \leq \theta.$$

Proof  $\varphi(t) = \langle F'(x+t(y-x)), y-x \rangle \quad t \in [0, 1]$

$$\varphi(0) \leq \theta$$

$\varphi(0) \leq 0 \Rightarrow$  trivial. Assume  $\varphi(0) > 0$ .

$$\varphi'(t) = \langle F''(x+t(y-x))(y-x), y-x \rangle$$

$$\geq \frac{1}{\theta} \langle F'(x+t(y-x)), y-x \rangle^2 = \frac{1}{\theta} \varphi(t)^2$$

•  $\varphi \nearrow$ .  $\varphi(t) \geq \varphi(0) \quad \forall t \in [0, 1]$ .

$$\bullet \frac{d}{dt} \left[ -\frac{1}{\varphi(t)} \right] = \frac{\varphi'(t)}{\varphi(t)^2} \geq \frac{1}{\theta}$$

$$\frac{1}{\varphi(0)} \geq \frac{1}{\varphi(0)} - \frac{1}{\varphi(1)} = \int_0^1 \frac{d}{dt} \left[ -\frac{1}{\varphi(t)} \right] \geq \frac{1}{\theta} \Rightarrow \varphi(0) \leq \theta. \quad \square$$

## Newton's step on Perturbed Problem.

$$x_t^* = \operatorname{argmin} [f_t(x) = t \langle c, x \rangle + F(x)]$$

Assume  $x \approx x_t^*$

$$\|f_t'(x)\|_x = \langle f_t'(x), F''(x)^{-1} f_t'(x) \rangle^{1/2} \leq \delta = \frac{1}{10}.$$

Update time:  $t_+ = t + \frac{\gamma}{\|c\|_x}$ ,  $\gamma = \frac{1}{10}$

$$\|c\|_x = \langle c, F''(x)^{-1} c \rangle^{1/2}$$

New objective:  $f_{t_+}(y)$

$$x_+ = x - F''(x)^{-1} (t_+ c + F'(x))$$

Theorem

$$\|f_{t_+}'(x_+)\|_{x_+} \leq \delta = \frac{1}{10}.$$