

→ We assume that the polytope  $P$  is **bounded**, full-dimensional & non-empty.

Initialization: goal: find  $\eta_0 > 0$  &  $x_0 \in \text{int}(P)$  s.t.

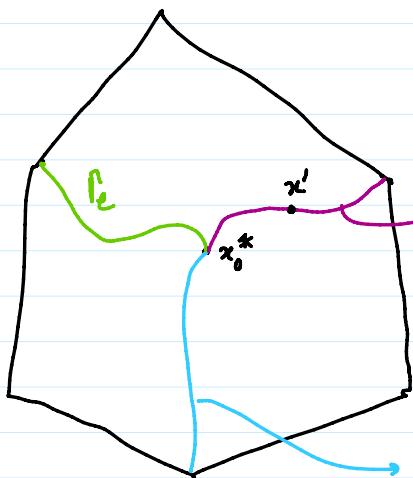
$$\|N_{\eta_0}(x_0)\|_{x_0} \leq 1/6$$

Recall:  $N_{\eta_0}(x_0) := -H_F(x_0)^{-1} \nabla f_{\eta_0}(x_0) = -H_F(x_0)^{-1} (\eta_0 c + \nabla F(x_0))$

$\eta_0$  will be  $\geq 2^{-\tilde{\Theta}(nL)}$  ( $\tilde{\Theta}(\cdot)$  hides logarithmic factors in various parameters).

First let us assume that we have access to an interior point  $x' \in P$ . In fact, we will assume something stronger. We will assume that we have an  $x' \in P$  s.t.

$$\langle a_i, x' \rangle \leq b_i - 2^{-\tilde{\Theta}(nL)} \quad \forall i \in [m]$$



$x_0^*$ : minimizer of  $F$   
 $x_\eta^*$ : minimizer of  $f_\eta = \eta \langle c, x \rangle + F(x)$

$$\Gamma_C = \{x_\eta^* : \eta > 0\}$$

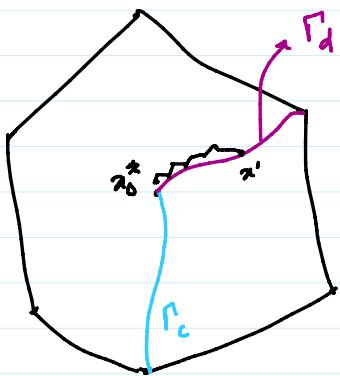
Observation: Each  $x' \in \text{int}(P)$  lies on some central path!

Df:  $d := -\nabla F(x')$ . Then  $x'$  lies on  $\Gamma_d$ .

$$f(x) = \langle d, x \rangle + F(x)$$

Then  $\nabla f(x') = d + \nabla F(x') = 0$ . Hence  $x'$  is the minimizer of  $f$ .

Then  $\nabla f(x^*) = d + \nabla F(x) = 0$ . Hence  $x^*$  is the minimizer of  $f$ .



Strategy: Start from  $x^*$  & traverse the central path  $\Gamma_c$  in reverse. Once we get close to  $x_0^*$ , then will close to the central path  $\Gamma_c$ .

$$(s_i(x) = b_i - \langle a_i, x \rangle)$$

Lemma (Following the central path  $\Gamma_d$  in reverse): Suppose  $x^*$  is a point s.t.  $s_i(x^*) \geq \beta \max_{x \in P} s_i(x) \forall i \in [m]$ .

$$N'_\eta(x) := -H_F(x)^{-1} (\eta d + \nabla F(x))$$

$\downarrow$

$$- \nabla F(x)$$

Then whenever  $\eta \leq \frac{\beta}{24\sqrt{m}}$  &  $\|N'_\eta(x)\|_x \leq \frac{1}{24}$ , it holds that  $\|H_F(x)^{-1} \nabla F(x)\|_x \leq \frac{1}{42}$ .

Proof: Use triangle inequality for the local norm

$$\begin{aligned} \|H_F(x)^{-1} \nabla F(x)\|_x &\leq \|N'_\eta(x)\|_x + \eta \|H_F(x)^{-1} d\|_x \\ &\leq \frac{1}{24} + \eta \boxed{\|H_F(x)^{-1} \nabla F(x)\|_x} \end{aligned}$$

$$\|H_F(x)^{-1} \nabla F(x)\|_x^2 = \langle \nabla F(x), H_F(x)^{-1} \nabla F(x) \rangle$$

Let  $A$  be an  $m \times n$  matrix whose rows are  $a_i$ 's.

$$A = \begin{pmatrix} -a_1 & \\ -a_2 & \\ \vdots & \\ -a_m & \end{pmatrix}$$

Let  $S_x$  denote the  $m \times m$  diagonal matrix that contains the slack values for  $x$ .

$$S_x = \begin{pmatrix} s_1(x) & & & \\ & \ddots & & 0 \\ 0 & & \ddots & \\ & & & s_m(x) \end{pmatrix}$$

$$\text{Then } H_F(x) = A^T S_x^{-1} A, \quad \nabla F(x) = A^T S_x^{-1} \mathbf{1}$$

$$\nabla F(x') = A^T S_{x'}^{-1} \mathbf{1}$$

$\mathbf{1}$ : all 1's vector ( $m \times 1$ ) .

$$\text{Then } \langle \nabla F(x'), H_F(x)^{-1} \nabla F(x') \rangle$$

$$= \mathbf{1}^T S_{x'}^{-1} A (A^T S_x^{-1} A)^{-1} A^T S_x \mathbf{1}$$

$$= \mathbf{1}^T S_{x'}^{-1} S_x S_x^{-1} A (A^T S_x^{-1} A)^{-1} A^T S_x^{-1} S_x S_x^{-1} \mathbf{1}$$

$$\text{where } v = S_x S_x^{-1} \mathbf{1}, \quad \Pi = S_x^{-1} A (A^T S_x^{-1} A)^{-1} A^T S_x^{-1}$$

$$\text{Let } B = S_x^{-1} A, \quad \text{then } \Pi = B \underbrace{(B^T B)^{-1}}_{\text{projection matrices}} B^T$$

$$\Pi^2 = B (B^T B)^{-1} \cancel{B^T} B \cancel{(B^T B)^{-1}} B^T$$

$$= B (B^T B)^{-1} B^T = \Pi$$

$$\Rightarrow v^T \Pi v = \|\Pi v\|_2^2 \leq \|v\|_2^2 \quad \text{since } \Pi \text{ is a projection matrix}$$

$$\|v\|^2 = \sum s_i(x)^2 \leq m$$

$$\|v\|_2^2 = \sum_{i=1}^m \frac{s_i(x)^2}{s_i(x')^2} \leq \frac{m}{\beta^2}$$

$$\Rightarrow \|H_F(x)^{-1} \nabla F(x)\|_x \leq \frac{\sqrt{m}}{\beta}$$

$$\begin{aligned} \Rightarrow \|H_F(x)^{-1} \nabla F(x)\|_x &\leq \frac{1}{24} + \eta \frac{\sqrt{m}}{\beta} \\ &\leq \frac{1}{24} + \frac{1}{24} = \frac{1}{12} \end{aligned}$$

Lemma (Switching the central path): Suppose  $x \in \text{int}(\rho)$

$$\text{s.t. } \|H_F(x)^{-1} \nabla F(x)\|_x \leq \frac{1}{12}. \quad \& \quad \eta_0 \text{ is s.t.}$$

$$\eta_0 \leq \frac{1}{12} \cdot \frac{1}{\langle c, x - x^* \rangle} \Rightarrow \|N_{\eta_0}(x)\|_x \leq \frac{1}{6}$$

$$\begin{aligned} \text{Proof: } \|N_{\eta_0}(x)\|_x &= \|H_F(x)^{-1} (\eta_0 c + \nabla F(x))\|_x \\ &\leq \eta_0 \|H_F(x)^{-1} c\|_x + \|H_F(x)^{-1} \nabla F(x)\|_x \\ &\leq \eta_0 \circled{ \|H_F(x)^{-1} c\|_x } + \frac{1}{12} \end{aligned}$$

Let  $c_x := H_F(x)^{-1} c$ . Then by the Dikin's Ellipsoid

$$\text{property, } x - \frac{c_x}{\|c_x\|_x} \in \rho$$

$$\Rightarrow \langle c, x - \frac{c_x}{\|c_x\|_x} \rangle \geq \langle c, x^* \rangle$$

$$\Rightarrow \langle c, x - x^* \rangle \geq 0, \quad \left\langle c, \frac{c_x}{\|c_x\|_x} \right\rangle = \frac{\|c_x\|_x^2}{\|c_x\|_x} \\ = \|c_x\|_x$$

$$\leq \eta_0 \langle c, x - x^* \rangle + \frac{1}{12}$$

$$\leq \frac{1}{12} + \frac{1}{12} = 1/6.$$

Lemma (Finding a starting point given an interior point):

There is an algorithm that given  $x'$  s.t.  $s_i(x') \geq \beta \max_{x \in P} s_i(x) + \epsilon$ ,

outputs a point  $x_0 \in \text{int}(P)$  &  $\eta_0 > 0$  s.t.

$$\|N_{\eta_0}(x_0)\|_{x_0} \leq 1/6.$$

$$\text{And also } \eta_0 \geq \frac{1}{\|c\|_2 \text{diam}(P)} \geq 2^{-\tilde{\Omega}(nL)}.$$

Running time:  $\text{poly}(n, m, L, \log(\frac{1}{\beta}))$

Proof: 1. Start by running the IPM in reverse for the central path  $P_d$ ,  $d := -\nabla F(x')$ .

Initially  $\eta = 1$  &  $\|N_1(x')\|_{x'} = 0 \leq 1/6$ .

2. Decrease  $\eta$  till it becomes less than

$$\eta_0 := \min \left\{ \frac{\beta}{2\sqrt{m}}, \frac{1}{\|c\|_2 \text{diam}(P)} \right\}$$

mult. decrease by a factor  $(1 - \frac{1}{2})$

↓ multiplicative decrease by a factor  $(1 - \frac{1}{20\sqrt{m}})$  at each step.

At this point, we get that  $\|N'_{\eta_0}(x)\|_x \leq \frac{1}{6}$ .

3. Do a couple of Newton iterations to get  $y$  s.t.

$$\|N'_{\eta_0}(y)\|_y \leq \frac{1}{24}.$$

4. From the first lemma, we get that

$$\|H_F(y)^{-1} \nabla F(y)\|_y \leq \frac{1}{12}$$

5. Since  $\langle c, y - x^* \rangle \leq \|c\|_2 \|y - x^*\|_2$   
 $\leq \|c\|_2 \text{diam}(\rho)$

$$\eta_0 \leq \frac{1}{\|c\|_2 \text{diam}(\rho)} \leq \frac{1}{\langle c, y - x^* \rangle}.$$

Hence we can apply the second lemma to get

$$\|N'_{\eta_0}(y)\|_y \leq \frac{1}{6}.$$

Exercise:  $\|c\|_2, \text{diam}(\rho) \leq 2^{\tilde{O}(nL)}$ .

Now how to find an interior point  $x^* \in P$ ?

Solution: Look at a different polytope  $P' \subseteq \mathbb{R}^{n+1}$

$$P' = \left\{ x \in \mathbb{R}^n, t \in \mathbb{R} \text{ s.t. } \langle a_i, x \rangle \leq b_i + t \quad \forall i \in [m] \right. \\ \left. -c \leq t \leq c \right\}$$

Take  $c = 2 + \sum_i |b_i|$ . Then the point-

$x' = 0, t' = 1 + \sum_i |b_i|$  is a point in  $\text{int}(P')$

& slack of each constraint  $\geq 1$ .

use this to solve the following linear program

$$\min_{(x,t) \in P'} t$$

suppose  $(x^*, t^*)$  is the optimal soln.

Then  $\langle a_i, x^* \rangle \leq b_i + t^* \quad \forall i$

$$t^* \leq -2^{-\tilde{\Omega}(nL)}$$

Exercise :

Solving the above program to precision/error  $2^{-\tilde{\Omega}(nL)}$

will give us a point  $x' \in P$  s.t.

$$\langle a_i, x' \rangle \leq b - 2^{-\tilde{\Omega}(nL)} \quad \forall i$$

Exercise : If  $P$  is bounded, full-dimensional & non-empty,  
then so is  $P'$ .

Alternate :

$$\min_{x \in P} \langle c, x \rangle$$

reduce to the following feasibility problem

$$\text{find a point in } P \cap \{x : \langle c, x \rangle \leq \gamma\}$$

& do a binary search on  $\mathcal{Y}$ .



### Self-concordant barrier functions

Definition (Self-concordant barrier function) :

Let  $K \subseteq \mathbb{R}^m$  be a convex set & let  $f: \text{int}(K) \rightarrow \mathbb{R}$  be

a thrice differentiable function. We say that  $F$  is a self-concordant barrier function with parameter  $\nu$  if it satisfies the following properties:

1. (Barrier):  $f(x) \rightarrow \infty$  as  $x$  approaches the boundary of  $K$ .
2. (Convexity):  $F$  is strictly convex.
3. (Complexity parameter):  $\|N_F(x)\|_x^2 = \langle \nabla F(x), H_F(x)^{-1} \nabla F(x) \rangle \leq \nu \quad \forall x \in \text{int}(K)$ .

4. (Self-concordance):  $\forall x \in \text{int}(K) \text{ & } \forall h$

$$\left| \frac{d^3}{dt^3} f(x+th) \Big|_{t=0} \right| \leq 2 \left( \frac{d^2}{dt^2} f(x+th) \Big|_{t=0} \right)^{3/2}$$

The factor of  $3/2$  in the exponent ensures scale-invariance i.e. if  $F(x)$  is self-concordant, then so is

$$F(\lambda x) \text{ for any } \lambda \in \mathbb{R}.$$

The logarithmic barrier function satisfies the above properties

with  $\nu = m$ .

Exercise: A self-concordant function  $F$  satisfies the NL condition

Exercise: A self-concordant function  $F$  satisfies the NL condition with  $S_0 = \frac{1}{6}$ .

Theorem: Let  $F$  be a self-concordant barrier function on a convex set  $K \subseteq \mathbb{R}^n$  with complexity parameter  $\nu$ . Then there is a IPM that after  $O(\sqrt{\nu} \log(\frac{\nu}{\varepsilon \eta_0}))$  iterations outputs  $\hat{x} \in K$  s.t.  $\langle c, \hat{x} \rangle - \langle c, x^* \rangle \leq \varepsilon$ . Each iteration requires solving a linear system of the kind  $H_F(x) y = z$  ( $y$  are the unknowns).

Search for efficient barrier functions:

$$P = \{x \in \mathbb{R}^n : \langle a_i, x \rangle \leq b_i \forall i \in [m]\}.$$

The logarithmic barrier has complexity parameter  $m$ .

Can we make it  $O(1)$ ?

Lower bound: Any self-concordant barrier function for  $K = [0, 1]^n$  has complexity parameter at least  $n$ .

Every convex set  $K \subseteq \mathbb{R}^n$  has a self-concordant barrier function with complexity parameter  $O(n)$ .

↓  
Constructions non-explicit

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↓  
Lee-Sidford constructed an explicit barrier function for  
polytopes which has complexity parameter  $O(n)$ .

[ ↓  
reduce linear programs to solving  $\tilde{O}(n)$  linear  
systems ]