

Local Convergence of Newton's Method

$f: Q \rightarrow \mathbb{R}$ convex, differentiable $Q \subseteq \mathbb{R}^n$ - open convex set

Main Lemma let f be self-concordant with $M \geq 0$.

$x \in Q$. let $y \in Q$ belong to Dikin's ellipsoid:

$$y \in \mathcal{E}_x = \left\{ y : \|y - x\|_x < \frac{2}{M} \right\}$$

Then, the Hessian is stable:

$$\eta := \|y - x\|_x < \frac{2}{M}$$

$$\left(1 - \frac{M}{2}\eta\right)^2 f''(x) \preceq f''(y) \preceq \left(1 - \frac{M}{2}\eta\right)^{-2} f''(x)$$

Newton's step

$$x^+ = x - f''(x)^{-1} f'(x) \Leftrightarrow f'(x) + f''(x)(x^+ - x) = 0$$

Why $x^+ \in Q$?

Newton's Decrement:

$$\begin{aligned} \lambda(x) &:= \|f'(x)\|_x \equiv \underbrace{\langle f'(x), f''(x)^{-1} f'(x) \rangle}^{\text{Dual Norm}}^{1/2} = \|f''(x)^{-1/2} f'(x)\|_2 = \\ &= \|x^+ - x\|_x \equiv \langle f''(x)(x^+ - x), x^+ - x \rangle^{1/2} \\ &= \langle x - x^+, f'(x) \rangle^{1/2} \end{aligned}$$

Remark: $\lambda(x^+) = 0$.

Theorem

let $\lambda(x) \leq \frac{1}{M}$. Then

$$M\eta \leq 1$$

$$\lambda(x^+) \leq M \lambda(x)^2$$

Region of quadratic convergence.

Proof

Denote $\eta := \lambda(x) \equiv \|x^+ - x\|_x$

$$\lambda(x^+) = \|f'(x^+)\|_{x^+} \leq \frac{1}{1 - \frac{M}{2}\eta} \|f'(x^+)\|_x$$

$$f'(x^+) = f'(x^+) - f'(x) - f''(x)(x^+ - x) = (G - H)(x^+ - x)$$

where $H = f''(x)$

Fundamental
theorem
of calculus

$$G = \int_0^1 f''(x + t(x^+ - x)) dt$$

$$\begin{aligned} \underline{\|f'(x^+)\|_x} &= \|(G - H)(x^+ - x)\|_x = \|H^{-1/2}(G - H)(x^+ - x)\|_2 = \\ &= \|H^{-1/2}(G - H)H^{-1/2}H^{1/2}(x^+ - x)\|_2 \leq \\ &\leq \underbrace{\|H^{-1/2}(G - H)H^{-1/2}\|_2}_{\leq ?} \cdot \underbrace{\|H^{1/2}(x^+ - x)\|_2}_{= r} \end{aligned}$$

$$G = \int_0^1 f''(x + \tau(x^+ - x)) d\tau \geq H \cdot \int_0^1 (1 - \tau \frac{M}{2} r)^2 d\tau = H \cdot \left(\underline{1 - \frac{Mr}{2}} + \frac{1}{12} M^2 r^2 \right)$$

$$H^{-1/2}(G - H)H^{-1/2} \geq I \cdot \frac{Mr}{2} \left(\frac{Mr}{6} - 1 \right)$$

$$G = \int_0^1 f''(x + \tau(x^+ - x)) d\tau \leq H \cdot \int_0^1 \frac{d\tau}{(1 - \tau \frac{Mr}{2})^2} = H \cdot \left(\frac{1}{1 - \frac{Mr}{2}} \right)$$

$$H^{-1/2}(G - H)H^{-1/2} \leq I \cdot \left[\frac{1}{1 - \frac{Mr}{2}} - 1 \right] = I \cdot \frac{Mr/2}{1 - Mr/2}$$

$$\|H^{-1/2}(G - H)H^{-1/2}\|_2 \leq \frac{Mr}{2} \max \left\{ \frac{1}{1 - \frac{Mr}{2}}, 1 - \frac{Mr}{6} \right\} \leq \frac{Mr/2}{1 - Mr/2}$$

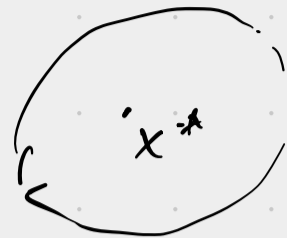
$$\lambda(x^+) \leq \frac{M}{2 - Mr} r^2 = \frac{M}{2 - M\lambda(x)} \lambda(x)^2 \leq M\lambda(x)^2 \quad \square$$

Discussion

1. The region of local quadratic convergence for Newton's method:

$$\lambda(x)^2 = \langle f'(x), f''(x)^{-1} f'(x) \rangle < \frac{1}{M^2}$$

2. Function is Quadratic: $M=0$. $x^+ = x^*$.



3. Standard Self-Concordant Function: $M=2$.

4. Fix Euclidean norm:

$$\left. \begin{array}{l} \cdot \text{Hessian is Lipschitz: } \|f''(x)\| \leq L \\ \cdot \text{Strongly convex: } f''(x) \succeq \mu I \end{array} \right\} \Rightarrow M = \frac{L}{\mu^{3/2}}$$

Strict local minimum: $f''(x^*) > 0$

$$\langle f'(x), f''(x)^{-1} f'(x) \rangle = \lambda(x)^2 \leq \frac{1}{\mu} \|f'(x)\|_2^2$$

Sufficient condition:

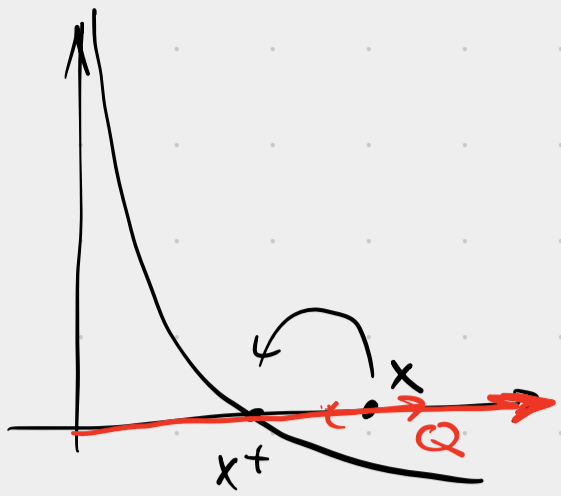
Corollary let $\|f'(x)\|_2 < \frac{\mu^2}{L}$ Then

$$\lambda(x) \leq \frac{1}{\sqrt{\mu}} \|f'(x)\|_2 \leq \frac{\mu^{3/2}}{L} = \frac{1}{M} \Rightarrow$$

x is in a region of quadratic convergence of Newton's method.

Quadratic function:

$$f(y) = \frac{1}{2} \langle Ay, y \rangle - \langle b, y \rangle \equiv f(x) + \langle f'(x), y-x \rangle + \frac{1}{2} \langle f''(x)(y-x), y-x \rangle$$



We use for $Q \subseteq \mathbb{R}^n$
 - "the largest possible set"

$f: Q \rightarrow \mathbb{R}$ "a natural domain" of f .

- $Q = \text{dom} f$
- f is closed convex function.

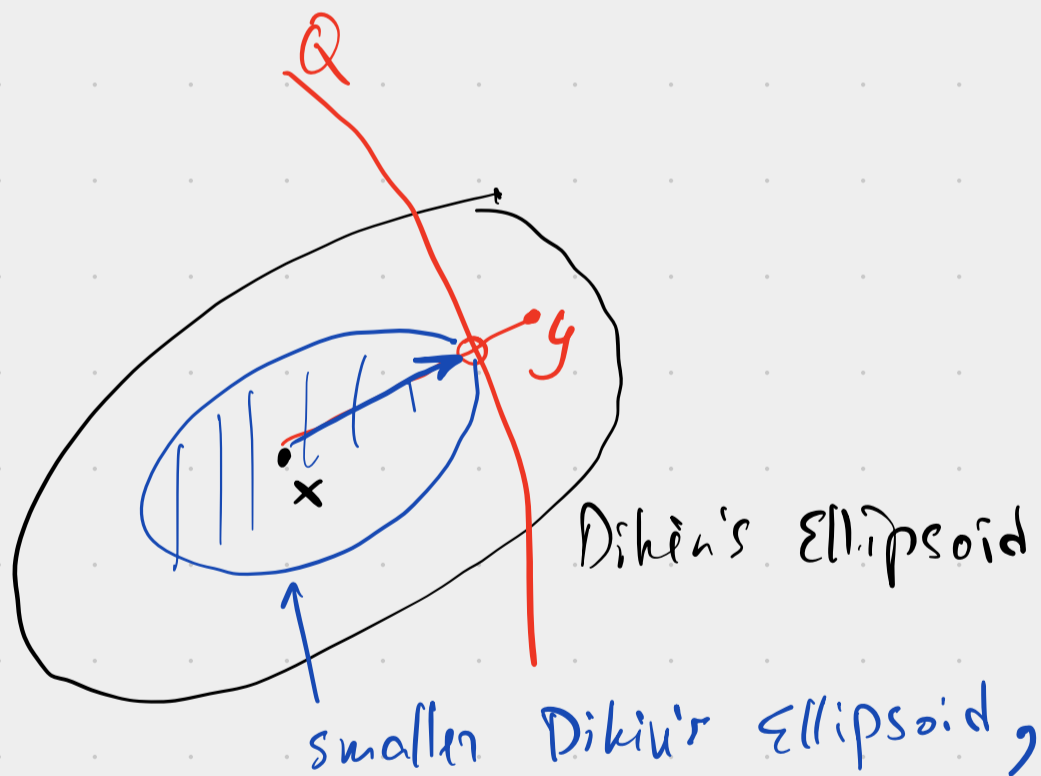
• For any $x_n \rightarrow \partial Q$ $f(x_n) \rightarrow +\infty$

Proposition let $f: Q \rightarrow \mathbb{R}$ be self-concordant, Q - natural domain.

Then, $x \in Q$:

$$E_x = \left\{ y : \|y - x\|_x < \frac{2}{M} \right\} \subseteq Q.$$

Proof sketch



Main lemma:

- Hessians are bounded
- f is bounded. (?!)

□

We proved: locally, $\lambda(x_k) \rightarrow 0, k \rightarrow +\infty$

$$f(x_k) - f^*$$

$$\|x_k - x^*\|_{x_k}$$

$$\|x_k - x^*\|_{x^*}$$

locally, they are related.

Proposition Assume $\lambda(x) < \frac{2}{M}$ Then

$$\|x - x^*\|_x \leq \frac{\lambda(x)}{1 - \frac{M}{2}\lambda(x)}.$$

$$\min_{x \in \Omega} f(x)$$

$$x \in \Omega$$

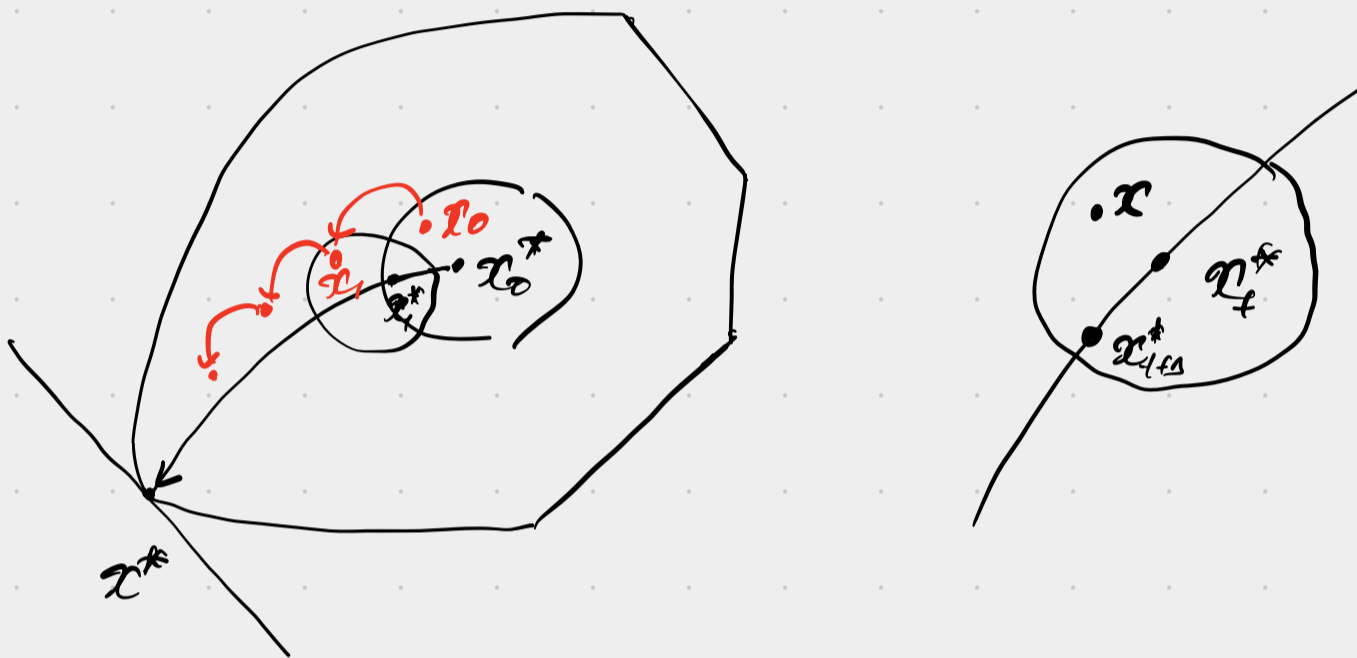
$$\Omega \subseteq \text{dom} f = Q$$

A similar result can be proved for other accuracy measures as well.

Path-Following Scheme

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

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



Idea: Introduce a "barrier function" $F: Q \rightarrow \mathbb{R}$
- helps to remain in the set.

We consider a family problems, $t \geq 0$:

$$f_t(x) = t \langle c, x \rangle + F(x)$$

Denote

$$x_t^* = \operatorname{argmin}_{x \in Q} f_t(x) \equiv \operatorname{argmin}_x f_t(x)$$

$$tc + F'(x_t^*) = 0$$

x_t^* - "central path"

$t=0$: $x_0^* = \operatorname{argmin}_x F(x)$ - an analytic center of Q .

$t \rightarrow +\infty$: $x_t^* \rightarrow x^*$