

# LP and its potential function

We start from the simple case of simplex-constrained QP

$$\begin{aligned} \min \quad & f(x) = \frac{1}{2} \|Ax\|_2^2 \\ \text{s.t.} \quad & e^T x = 1, \quad x \geq 0. \end{aligned}$$

Define the potential function and its gradient and Hessian

$$\phi(x) = \rho \log(f(x)) - \sum_{i=1}^n \log x_i,$$

$$\nabla \phi(x) = \frac{\rho \nabla f(x)}{f(x)} - X^{-1} e,$$

$$\nabla^2 \phi(x) = H = -\frac{\rho}{f(x)^2} \nabla f(x) \nabla f(x)^T + \rho \frac{A^T A}{f(x)} + X^{-2}.$$

# Full-Dimension DRSOM

Assume we solve the full-dimension problem in each iterate:

$$\begin{aligned} \min_d \quad & \nabla\phi(x)^T d + \frac{1}{2} d^T H d \\ \text{s.t.} \quad & e^T d = 0, \quad \|X^{-1}d\|_2 \leq \beta. \end{aligned} \tag{1}$$

## Theorem 1

*By choosing  $\beta \leq c\sqrt{f(x)}$  for some  $c > 0$ , there exists some  $M > 0$  such that*

$$\phi(x + d) - \phi(x) \leq \nabla\phi(x)^T d + \frac{1}{2} d^T H d + \frac{M\beta^3}{2f(x)^{\frac{3}{2}}} + \beta^3.$$

# Proof of Theorem 1

## Lemma 1

For any  $x > -1$ , we have

$$\left| \log(1+x) - x + \frac{x^2}{2} \right| \leq \frac{|x|^3}{3}.$$

According to Lemma 1, we have

$$\begin{aligned} \log(f(x+d)) - \log(f(x)) &= \log\left(1 + \frac{\nabla f(x)^T d}{f(x)} + \frac{\|Ad\|_2^2}{2f(x)}\right) \\ &\leq \frac{\nabla f(x)^T d}{f(x)} + \frac{\|Ad\|_2^2}{2f(x)} - \frac{d\nabla f(x)\nabla f(x)^T d}{2f(x)^2} + \frac{M\beta^3}{2f(x)^{\frac{3}{2}}} \end{aligned}$$

and

$$\begin{aligned} -\sum_{i=1}^n \log(x_i + d_i) + \sum_{i=1}^n \log(x_i) &= -\sum_{i=1}^n \log\left(1 + \frac{d_i}{x_i}\right) \\ &\leq -d^T X^{-1}e + d^T X^{-2}d + \sum_{i=1}^n \left| \frac{d_i}{x_i} \right|^3 \leq -d^T X^{-1}e + d^T X^{-2}d + \beta^3. \end{aligned}$$

# Optimality Condition

The problem (1) can be reformulated as

$$\begin{aligned} \min_d \quad & \nabla\phi(x)^T X d' + \frac{1}{2} d'^T Q d' \\ \text{s.t.} \quad & e^T X d' = 0, \quad \|d'\|_2 \leq \beta. \end{aligned} \tag{2}$$

where  $Q = X H X$ . Its optimality condition is

$$\begin{aligned} (Q + \mu I) d' &= -X(\nabla\phi(x) - \lambda e), \\ e^T X d' &= 0, \quad \|d'\|_2 \leq \beta \\ Q + \mu I &\succeq 0, \\ \mu &\geq 0, \quad \mu(\beta - \|d'\|_2) = 0. \end{aligned}$$

## Decrease in each step

Firstly,

$$\nabla\phi(x)^T X d' + \frac{1}{2} d'^T Q d' = -d'^T (Q + \mu I) d' + \frac{1}{2} d'^T Q d' \leq -\frac{\mu}{2} \|d'\|_2^2.$$

### Assumption 1 (Scaled Lipschitz)

*The scaled hessian  $Q = X H X$  is bounded by*

$$\|Q\|_2 \leq c f(x)^{-\frac{3}{4}}.$$

According to Ye(2005),

$$\|(Q + \mu I) d'\| = \|X(\nabla\phi(x) - \lambda e)\|_2 \geq 1.$$

By choosing  $\beta \leq \frac{f(x)^{\frac{3}{4}}}{2c}$ , then

$$1 \leq \|(Q + \mu I) d'\| \leq (\|Q\|_2 + \mu) \beta \leq \frac{1}{2} + \mu \beta$$

which implies  $\mu \geq 1/2\beta$ .

## Decrease in each step

Thus, if  $\beta \leq \frac{f(x)^{\frac{3}{4}}}{2c}$ , we have

$$\nabla \phi(x)^T X d' + \frac{1}{2} d'^T Q d' \leq -\frac{\beta}{4}.$$

By choosing  $\beta = O(f(x)^{\frac{3}{4}})$ , theorem 1 shows

$$\phi(x+d) - \phi(x) \leq -\frac{\beta}{4} + \frac{M\beta^3}{2f(x)^{\frac{3}{2}}} + \beta^3 \leq -c_0 f(x)^{\frac{3}{4}}.$$

If  $f(x^k) \geq \epsilon f(x^0)$  for  $k = 1, \dots, K$ , we must have

$$\rho \log\left(\frac{1}{\epsilon}\right) \geq \phi(x^0) - \phi(x^K) \geq \sum_{k=1}^K c_0 f(x^k) \geq c_0 f(x^0)^{\frac{3}{4}} \epsilon^{\frac{3}{4}} K$$

that is,

$$K \leq \frac{\rho / (c_0 f(x^0)^{\frac{3}{4}})}{\epsilon^{\frac{3}{4}}} \log\left(\frac{1}{\epsilon}\right).$$