

Interior Point Algos :

← say polytope.

Suppose $P = \{x : Ax \geq b\}$ want to $\min \text{ (something)} = C^T x$ st $x \in P$.

Constrained problem. Let's throw them away; replace by a "barrier"

$$\min_{\eta} f_{\eta}(x) = C^T x + \eta \cdot \sum_{i=1}^m \ln(1/a_i x - b_i)$$

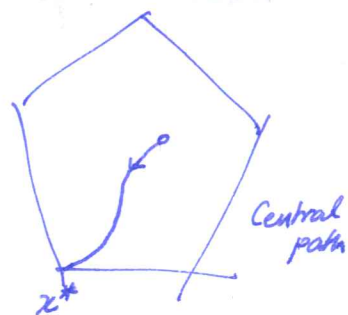
so as $a_i x \rightarrow b_i$ the function tends to ∞ . (Hence the function is making a barrier as we reach the boundary of P).

Hence as $\eta \rightarrow 0$, we need to assume that the polytope P has a non-empty interior.

Now the minimizer of $f_{\eta}(x)$ is called the "analytic center". Note it depends on η . But also on the representation of P . E.g. if you double the constraint # i then the function f_{η} and the associated analytic center changes.

Now consider the path as $\eta \rightarrow 0$. In that case this ~~path~~ x_{η}^* analytic center goes to the actual minimizer.

Algorithms that (try to) follow this path are called path-following methods. Idea: if you have a point x_{η} (close to x_{η}^*) then getting a point $x_{\eta'}$ (close to $x_{\eta'}^*$) is not that difficult. So keep moving "close to" the central path until very close to x^* , and then round much like in ellipsoid



OK. Let's consider an LP in equational form for the rest of the discussion.

$$\begin{aligned} \min \quad & C^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq 0. \end{aligned}$$

(primal)

$$\Leftrightarrow \begin{aligned} \max \quad & b^T y \\ \text{s.t.} \quad & A^T y \leq C \end{aligned}$$

(dual)

Also assume that both primal and dual are strictly feasible.
Both have solutions in their interiors.
technical requirement

To solve this again add the barrier to get

(2)

$$\min f_q(x) = c^T x - \eta \sum_{i=1}^n \ln x_i$$

$$Ax = b.$$

↑
advantage of equality form.

How to find the minimizer? Method of Lagrange Multipliers.

"Lagrangify" the constraints

$$\min f_q(x) - \sum_{i=1}^m y_i (a_i x - b_i) = f_q(x) - y^T (Ax - b).$$

At the minimizer we should have gradients = 0.

$$\text{gradient wrt } x : \nabla f_q(x) - y^T A = 0$$

$$\text{wrt } y : Ax - b = 0.$$

$$\text{What is } \nabla f_q(x) = c^T - \eta \left(\frac{1}{x_1}, \frac{1}{x_2}, \dots, \frac{1}{x_n} \right) \Rightarrow Ax = b$$

$$c^T - \eta \left(\frac{1}{x} \right)^T = y^T A$$

↑ rotation for $(\frac{1}{x_1}, \dots, \frac{1}{x_n})$

⇒

$$\begin{aligned} Ax &= b \\ A^T y + \bar{s} &= c \\ \bar{s} &= \eta \left(\frac{1}{x} \right) \quad \text{or} \quad \bar{s}_i x_i = \eta \quad \forall i \end{aligned}$$

↑
"slack"

Two comments:

at the minimum

① Lagrange multipliers show necessity of existence of y, s such that these conditions are satisfied. Need to show they are sufficient to capture the minimum, need to use the strict feasibility for that. See [Matuszewska-Gieratner] for details.

② Sp $\eta = 0$ then this is exactly complementary slackness!

$$c^T x = (y^T A x + s^T x) = y^T A x = y^T b \Rightarrow \text{primal} = \text{dual}!$$

Good. we have the system $Ax=b$ $A^T y + s = c$] maintain exactly
 $s \cdot x = \eta \mathbb{1}$ maintain approxiately.

In particular make sure that

$$\|s \circ x - \eta \mathbb{1}\|_2 \leq \theta \cdot \eta \quad \text{for } \boxed{\theta = 0.4}. \quad (*)$$

Alg.: ① Start with $s_0, x_0 > 0$ st $Ax_0 = b$
 y_0 $A^T y_0 + s_0 = c$

How? see books ...

and (*) is satisfied. Define $\eta_0 = \frac{\langle x_0, s_0 \rangle}{n}$ for η_0 .

② at each time t , find $\text{sol}^M(\Delta x, \Delta y, \Delta s)$ to

$$A \Delta x = 0$$

$$A^T \Delta y + \Delta s = 0$$

$$(s \circ \Delta x) + (x \circ \Delta s) = - (x \circ s)_t + \eta_t \mathbb{1}$$

where $\eta_t = \theta \cdot \eta_{t-1}$

$$\boxed{\sigma = (1 - \frac{\theta}{\sqrt{n}})}$$

set $x_t \leftarrow x_{t-1} + \Delta x$ etc.

↑ intuition: drop $(\Delta x \circ \Delta s)$ quadratic term.

Solvable by "necessary" part of Lagrange multi argument.

Claims: (1) $\forall t$, s_t, x_t, y_t satisfy $Ax=b$ $A^T y + s = c$. (By construction)

To prove! $\rightarrow$ (2) $\forall t$ s_t, x_t satisfy $\|x_t \circ s_t - \eta_t \mathbb{1}\| \leq \theta \eta_t$ (*)

(3) $\eta_t \leq \sigma^t \cdot \eta_0 \leq \sigma^t \cdot 2^{\theta(L)}$ where $L = \langle A \rangle + \langle b \rangle + \langle c \rangle$

$\Rightarrow$ after $t = O(\sqrt{n}L)$ rounds $\eta_t \leq 2^{-\theta(L)}$

(4) $\langle \Delta x, \Delta s \rangle = 0$ $\forall$ inst.

Pf: Indeed $\Delta s = -A^T \Delta y \Rightarrow \langle \Delta x, \Delta s \rangle = -\Delta x^T A^T \Delta y = 0$. $\square$

(5) $\eta_t = \frac{\langle x_t, s_t \rangle}{n}$

Pf: $\langle x_t, s_t \rangle = \langle \mathbb{1}, x_t \circ s_t \rangle = \langle \mathbb{1}, \eta_t \mathbb{1} - \Delta x_t \circ \Delta s_t \rangle = \eta_t \cdot n - 0$ (by 4) $\square$

(4)

(6) After $T = O(\frac{1}{\epsilon})$ steps almost optimal solⁿs.

$$\eta_t = \frac{1}{n} \langle x_t, s_t \rangle = \frac{1}{n} \langle x_t, -A^T y_t + c \rangle = \frac{1}{n} \langle c^T x_t - b^T y_t \rangle \\ = \frac{1}{n} (\text{duality gap})$$

$\Rightarrow x_t, y_t$ are both nearly optimal upto $2^{-O(t)}$ error.

Finally, remains to prove #2 above. Pf not particularly edifying, but here it is.

pf: $\|x_t \circ s_t - \eta_t \mathbb{1}\|_2^2 = \|\Delta x \circ \Delta s\|_2^2$ by construction. And while the inner product $\langle \Delta x \circ \Delta s \rangle = 0$ no control per se on per-entry.

Let $D = \text{diag}(\sqrt{x_i/s_i})$.

$x = x_t$
 $s = s_t$

then $\|\Delta x \circ \Delta s\| = \|(D^{-1} \Delta x) \circ (D \Delta s)\|$

$$\|a \circ b\| \leq 2^{3/2} \|a\| \|b\|$$

$$\leq 2^{-3/2} \|D^{-1} \Delta x + D \Delta s\|^2$$

$$= 2^{-3/2} \|\sqrt{\frac{1}{x_i} \circ \frac{1}{s_i}} (s \Delta x + x \Delta s)\|^2$$

$$= 2^{-3/2} \left\| \sqrt{\frac{1}{x} \circ \frac{1}{s}} (-x \circ s_{t-1} + \eta_t \mathbb{1}) \right\|^2$$

$$= 2^{-3/2} \sum_i \frac{(x_i s_i + \eta_t)^2}{x_i s_i} \leq 2^{-3/2} \frac{\|x \circ s_{t-1} - \eta_t \mathbb{1}\|^2}{\min x_i s_i}$$

but $x_i s_i \geq 1 - \theta \eta_{t-1}$ (by induction)

and $\|x_{t-1} \circ s_{t-1} - \eta_{t-1} \mathbb{1}\|^2$

$$= \|(x_{t-1} \circ s_{t-1} - \eta_{t-1} \mathbb{1}) + (1 - \sigma) \eta_{t-1} \mathbb{1}\|^2$$

$$= \|x_{t-1} \circ s_{t-1} - \eta_{t-1} \mathbb{1}\|^2 + (1 - \sigma)^2 \|\eta_{t-1} \mathbb{1}\|^2 + 0$$

$$\leq \theta^2 \eta_{t-1}^2 + (1 - \sigma)^2 \eta_{t-1}^2 \cdot \eta^2$$

$$\text{if } \sigma = 1 - \theta/n \Rightarrow 2\theta^2 \eta_{t-1}^2$$

And the $2^{-3/2}$ factor can give us $\leq \theta^2 \eta_t^2$.
strict

strict

(5)

Let's say the argument again.

Start with $x_0, s_0 > 0$. Define $\eta_0 = \frac{\langle x_0, s_0 \rangle}{n}$

each time x_t, s_t obtained by adding in $\Delta x, \Delta s$. Maintain $\eta_t = \frac{\langle x_t, s_t \rangle}{n}$ by fact 5.

Finally get x_t, y_t to have $c^T x_t - b^T y_t \leq \eta_t$ and stop.

What about feasibility?

Clearly $Ax = b$, $A^T y + s = c$ satisfied. But $x, s \geq 0$?

In fact maintain $x, s > 0$.

Start with such a $s_0 \in \mathbb{R}^n$. $x_0, s_0 > 0$.

Now claim that preserved at time t .

(a) if some $x_i s_i \leq 0$ then ℓ_2 -distance would give no η_t just from that coord, and hence violate (*). In fact $x_i s_i < \eta_t \times 0.6$ would do it.

(b) what if both x_i and $s_i < 0$ but $x_i s_i > 0$?

Can actually use the argument to show that not just (x_t, s_t) satisfies (*) but all points of form $(x_t + \alpha \Delta x, s_t + \alpha \Delta s)$ for $\alpha \in [0, 1]$ satisfy (*). So all is feasible.

