

COMPUTATIONAL METHODS IN OPTIMIZATION

December 4, 2023

Today: Final lecture on implementing IPMs

Next sessions: Lab/course project

- Tuesday Dec 5 3.30 - 6.45 pm
- Monday Dec 11 8.30 - 11.45 am
- Tuesday Dec 12 3.30 - 5 pm
- Thursday Dec 14 1.45 - 3.15 pm

IMPLEMENTING INTERIOR-POINT METHODS

↳ IPMs are well implemented when

- efficient linear algebra techniques are used
- the algorithm is tailored to the problem of interest
- the iterates remain in the interior of the primal-dual feasible set (central path neighborhood)

↳ Today: illustration of this for LPs (but this is even more important for conic programming in general)

① Linear algebra in IPMs for LPs

Consider a linear program in standard form and its dual:

$$(P) \quad \underset{x \in \mathbb{R}^n}{\text{minimize}} \quad c^T x \quad \text{s.t.} \quad Ax = b, \quad x \geq 0$$

$$(D) \quad \underset{\substack{y \in \mathbb{R}^m \\ s \in \mathbb{R}^n}}{\text{maximize}} \quad b^T y \quad \text{s.t.} \quad A^T y + s = c, \quad s \geq 0$$

IPMs: Start with (x^0, y^0, s^0) such that $x^0 > 0, s^0 > 0$

$$I_n = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \cdot \text{ For } k=0, 1, \dots$$

$$e = \begin{bmatrix} 1 \\ 1 \\ \vdots \end{bmatrix} \in \mathbb{R}^n$$

• Solve

$$\begin{bmatrix} A & 0 & 0 \\ 0 & A^T & I_m \\ S^k & 0 & X^k \end{bmatrix} \begin{bmatrix} \Delta x^k \\ \Delta y^k \\ \Delta s^k \end{bmatrix} = - \begin{bmatrix} Ax^k - b \\ A^T y^k + s^k - c \\ X^k S^k e \end{bmatrix}$$

• Set $(x^{k+1}, y^{k+1}, s^{k+1}) = (x^k, y^k, s^k) + \alpha^k (\Delta x^k, \Delta y^k, \Delta s^k)$
 where α^k is chosen so that $x^{k+1} > 0$ and $s^{k+1} > 0$
 (+ possibly additional conditions)

• Advanced variants: Replace $x^k s^k e$ by $x^k s^k e - \frac{\mu^k e}{\sigma^k}$, with $\mu^k = \frac{(x^k)^T s^k}{n}$
 and $\sigma^k > 0$

↳ The main cost of an IPM iteration is solving the linear system (size $(m+2n) \times (m+2n)$)

⇒ In practice, this system is never solved in the form above, but it is reduced to a smaller system of linear equations

Augmented system

(For simplicity, we drop the k superscripts)

$$\text{System} \quad \begin{matrix} \overset{m}{\leftarrow} & \overset{m}{\leftarrow} & \overset{m}{\leftarrow} \\ \begin{matrix} \downarrow m \\ \downarrow m \\ \downarrow m \end{matrix} \end{matrix} \begin{bmatrix} A & 0 & 0 \\ 0 & A^T & I_n \\ S & 0 & X \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \\ \Delta s \end{bmatrix} = \begin{bmatrix} r_b \\ r_c \\ r_{xs} \end{bmatrix} \begin{matrix} \downarrow m \\ \downarrow n \\ \downarrow n \end{matrix}$$

$$S = \text{diag}(s) = \begin{bmatrix} s_1 & & 0 \\ & \ddots & \\ 0 & & s_m \end{bmatrix}$$

$$r_b = b - Ax, \quad r_c = c - A^T y - s, \quad r_{xs} = -XSe$$

$$X = \text{diag}(x)$$

↳ The last n equations of the system give

$$S \Delta x + X \Delta s = r_{xs}$$

Needed for the IPM iteration

Since $x > 0$ and $s > 0$, S and X are invertible matrices, hence $S^{-1} = \begin{bmatrix} 1/s_1 & & 0 \\ & \ddots & \\ 0 & & 1/s_m \end{bmatrix}$ and $X^{-1} = \begin{bmatrix} 1/x_1 & & 0 \\ & \ddots & \\ 0 & & 1/x_n \end{bmatrix}$

are well-defined.

Thus, we can express Δs as a function of Δx as follows

$$\Delta s = X^{-1} r_{xs} - X^{-1} S \Delta x.$$

Substituting the expression for Δs in the main linear system, we reduce it to a system of size $m+n$ called the augmented system:

$$\begin{bmatrix} -\textcircled{H}^{-1} & A^T \\ A & 0 \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} = \begin{bmatrix} r_c - X^{-1} r_{xs} \\ r_b \end{bmatrix}$$

where $\textcircled{H} = X S^{-1} = \begin{bmatrix} x_1/s_1 & 0 \\ 0 & x_n/s_n \end{bmatrix}$ is diagonal

Idea: • If we solve the augmented system for $(\Delta x, \Delta y)$, we get Δs immediately by computing $\underbrace{X^{-1} r_{xs}}_{\text{diagonal matrix}} - \underbrace{X^{-1} S \Delta x}_{\text{diagonal matrix}}$

• The augmented system is of size $m+n$ whereas the original system was of size $m+2n$
 \Rightarrow the augmented system is in general easier to solve computationally

\rightarrow The augmented system approach applies beyond LP
(e.g. for QPs, the only difference with the approach above is that $-\textcircled{H}^{-1}$ term becomes $-Q - \textcircled{H}^{-1}$ where Q is the matrix representing the quadratic term in the objective)

Normal equations

↳ One step further than the augmented system, typically used for LPs but harder to use on more general classes of problems

↳ Starts from the augmented system

$$\begin{matrix} m+n \\ \uparrow \end{matrix} \begin{bmatrix} -\kappa^{-1} & A^T \\ A & 0 \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} = \begin{bmatrix} r_c - X^{-1} r_{xs} \\ r_b \end{bmatrix}$$

$\leftarrow n + m$

First n equations

$$-\kappa^{-1} \Delta x + A^T \Delta y = r_c - X^{-1} r_{xs}$$

$$\Leftrightarrow -\Delta x + \kappa A^T \Delta y = \kappa r_c - \kappa X^{-1} r_{xs}$$

$$\Leftrightarrow \Delta x = \kappa A^T \Delta y - \kappa r_c + \kappa X^{-1} r_{xs}$$

↳ Plugging the expression for Δx in the last m equations, we obtain

$$A \Delta x = r_b$$

$$\Leftrightarrow A \kappa A^T \Delta y - A \kappa r_c + A \kappa X^{-1} r_{xs} = r_b$$

$$\Leftrightarrow A \kappa A^T \Delta y = r_b + A \kappa r_c - A \kappa X^{-1} r_{xs}$$

$\in \mathbb{R}^{m \times m}$

"Normal equations"

Linear system of m equations with m unknowns

↳ Idea: If we solve the normal equations for Δy , we can then compute Δx (and thus Δs) without having to solve another linear system.

\Rightarrow The normal equations are typically easier to solve numerically than the augmented system

② Exploiting structure

Motivating (LP) example

$$\begin{aligned} &\text{minimize } c^T x \quad \text{s.t.} \quad Ax = b \\ &x \in \mathbb{R}^m \quad \quad \quad 0 \leq x \leq e \quad \quad e = \begin{bmatrix} 1 \\ \vdots \\ 1 \end{bmatrix} \in \mathbb{R}^m \end{aligned}$$

\hookrightarrow If we reformulate the problem in standard form, we would obtain a problem in $2m$ variables with $m + 3m$ dual variables.

$$\begin{aligned} &\text{minimize } c^T x \quad \text{s.t.} \quad m Ax = b \\ &x \in \mathbb{R}^m \quad \quad \quad m(x \geq 0) \\ &w \in \mathbb{R}^m \quad \quad \quad m(w = e - x) \quad (\Leftrightarrow x + w = e) \\ &\quad \quad \quad \quad \quad \quad m(w \geq 0) \end{aligned}$$

\Rightarrow At every iteration of IPM, we would need to solve a linear system of $2m + m + 3m = m + 5m$ equations.

\Rightarrow with the normal equations reformulation, we can reduce it to a system of $m + m$ equations ($m + 5m \rightarrow m + 3m$ augmented system)

\hookrightarrow Generic approach for LPs in standard form, does not use the structure of this specific problem $\rightarrow m + m$ normal equations)

\hookrightarrow We can reduce the size of the linear system to be solved by exploiting the particular structure of the problem

$$\begin{aligned} &\text{minimize } c^T x \quad \text{s.t.} \quad Ax = b \\ &x \in \mathbb{R}^n \\ &w \in \mathbb{R}^m \\ &x + w = e \\ &x \geq 0 \\ &w \geq 0 \end{aligned}$$

The dual of this problem is

$$\begin{aligned} &\text{maximize } b^T y \quad \text{s.t.} \quad A^T y + z + s = c \\ &y \in \mathbb{R}^m \\ &z \in \mathbb{R}^m \\ &s \in \mathbb{R}^m \\ &t \in \mathbb{R}^n \\ &z + t = 0 \\ &s \geq 0 \\ &t \geq 0 \end{aligned}$$

(Standard LP dual for minimize $\hat{c}^T \hat{x}$ s.t. $A^1 \hat{x} = \hat{b}$
 $\hat{x} \in \mathbb{R}^{\hat{m}}$ $\hat{x} \geq 0$)

$$\text{with } \hat{x} = \begin{bmatrix} x \\ w \end{bmatrix}, \hat{c} = \begin{bmatrix} c \\ 0 \end{bmatrix}, A^1 = \begin{bmatrix} A & 0 \\ I & I \end{bmatrix} \in \mathbb{R}^{(m+n) \times (n+m)}, \hat{b} = \begin{bmatrix} b \\ e \end{bmatrix} \in \mathbb{R}^{m+n}$$

Dual

$$\left(\begin{array}{l} \hat{y} = \begin{bmatrix} y \\ z \end{bmatrix} \\ \hat{s} = \begin{bmatrix} s \\ t \end{bmatrix} \end{array} \leftarrow \begin{array}{l} \text{maximize } \hat{b}^T \hat{y} \\ \hat{y} \in \mathbb{R}^{m+n} \\ \hat{s} \in \mathbb{R}^{2n} \end{array} \text{ s.t. } \begin{array}{l} A^1 \hat{y} + \hat{s} = \hat{c} \\ \hat{s} \geq 0 \end{array} \right)$$

→ In the dual, the variables z can be eliminated since $z = -t$

→ we end up with a dual of the form

$$\begin{aligned} &\text{maximize } b^T y \quad \text{s.t.} \quad A^T y - t + s = c \\ &y \in \mathbb{R}^m \\ &s, t \in \mathbb{R}^n \\ &s \geq 0 \\ &t \geq 0 \end{aligned}$$

The reasoning from (1) can be applied to reduce the IDP iteration to a solve of a system of m linear equations (in y)

$$\begin{bmatrix} \dots \\ \dots \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta w \\ \Delta y \\ \Delta s \\ \Delta t \end{bmatrix} = \dots$$

\Downarrow

$$A \widehat{H}^{-1} A^T \Delta y = \dots$$

$$\widehat{H} = X^{-1} S + W^{-1} T$$

$$W = \text{diag}(w)$$

$$T = \text{diag}(t)$$