

# COMPUTATIONAL METHODS IN OPTIMIZATION

October 9, 2023

Today: End of "crash course" on interior-point methods  
Next session: Monday, October 23

# ① Path-following IPMs

Setup: linear programming

$$(P) \quad \begin{array}{ll} \text{minimize}_{x \in \mathbb{R}^m} & c^T x \\ & \text{s.t. } Ax = b \\ & \quad x \geq 0 \end{array}$$

$$(D) \quad \begin{array}{ll} \text{maximize}_{y \in \mathbb{R}^m} & b^T y \\ & \text{s.t. } A^T y + s = c \\ & \quad s \geq 0 \end{array}$$

$$A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m, c \in \mathbb{R}^n$$

↳ Primal-dual IPMs generate a sequence of primal-dual iterates  $\{(x^k, y^k, s^k)\}_{k \in \mathbb{N}}$  such that

- $x^k \geq 0 \quad \forall k$
- $s^k \geq 0$
- $\frac{(x^k)^T (s^k)}{m} \xrightarrow[k \rightarrow \infty]{\rightarrow} 0$

↓  
"Interior points"

↳ Generic algorithm:

- Choose  $(x^0, y^0, s^0)$
- Given  $(x^k, y^k, s^k)$ , compute

$$\begin{bmatrix} x^{k+1} \\ y^{k+1} \\ s^{k+1} \end{bmatrix} = \begin{bmatrix} x^k \\ y^k \\ s^k \end{bmatrix} + \alpha^k \begin{bmatrix} \Delta x^k \\ \Delta y^k \\ \Delta s^k \end{bmatrix}$$

$\alpha^k$  "Search direction"  
 $\Delta x^k, \Delta y^k, \Delta s^k$  "Stepsize"  $> 0$

↳ Assumption (for today):

$(x^0, y^0, s^0)$  is primal-dual **strictly feasible**

i.e.

$$Ax^0 = b$$

$$x^0 \geq 0$$

$$A^T y^0 + s^0 = c$$

$$s^0 \geq 0$$

- Infeasible IPMs do not require this assumption (and typically work better than methods that do!)

- Homogeneous self-dual IPNs satisfy the assumption automatically by reformulating the problem (The solver MOSEK uses this)

## Computing the search direction

Approach: Use the KKT conditions for (P) and (D)

$(x^*, y^*, s^*)$  is a primal-dual solution ( $\Rightarrow$ )

$$\left\{ \begin{array}{l} Ax^* - b = 0 \\ A^T y^* + s^* - c = 0 \\ x^* \geq 0 \\ s^* \geq 0 \\ x_i^* s_i^* = 0 \quad \forall i=1..n \end{array} \right.$$

- In a simplex method, the exact solution is obtained by looking at  $(x, y, s)$  that satisfy  $x_i s_i = 0 \quad \forall i=1..n$  ( $\Rightarrow$  fix some  $x_i$  and  $s_i$  to be 0)
- In an IPM, one considers only approximate solutions  $(x^k, y^k, s^k)$  such that  $x^k > 0$  ( $\Rightarrow x_i^k s_i^k > 0$ ) that get closer and closer to an exact solution

↳ An exact solution satisfies (in particular)

Nonlinear system of equations

2n+m equations  
2n+m variables  $(x^*, y^*, s^*)$

$$\left\{ \begin{array}{l} Ax^* - b = 0 \\ A^T y^* + s^* - c = 0 \\ x^* s^* = 0 \end{array} \right.$$

$$x^* = \begin{bmatrix} x_1^* \\ \vdots \\ x_n^* \end{bmatrix}$$

$$s^* = \begin{bmatrix} s_1^* \\ \vdots \\ s_m^* \end{bmatrix}$$

Given  $(x^k, y^k, s^k)$ , one can apply Newton's method for nonlinear equations starting from  $(x^k, y^k, s^k)$  to obtain a direction towards a better point in the sense of the nonlinear system  $Ax - b = 0$

$$A^T y + s - c = 0$$

$$x^k s^k e = 0$$

→ In practice, we apply 1 iteration of Newton's method, which corresponds to solving the linear system

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

$\underbrace{\qquad\qquad\qquad}_{\text{unknowns}}$

↑  
Nonlinear system  
at  $(x^k, y^k, s^k)$

→ If  $(x^k, y^k, s^k)$  is primal-dual strictly feasible, the right-hand side is  $\begin{bmatrix} 0 \\ 0 \\ -x^k s^k e \end{bmatrix}$  with the last 2 coefficients being nonzero.

Two issues with this direction

- Moving in the direction might lead to negative coefficients for  $x^{k+1}$  and  $s^{k+1}$
- Newton's method does not always converge

→ Because of these issues, one cannot simply do

$$\begin{bmatrix} x^{k+1} \\ y^{k+1} \\ s^{k+1} \end{bmatrix} = \begin{bmatrix} x^k \\ y^k \\ s^k \end{bmatrix} + \begin{bmatrix} \Delta x^k \\ \Delta y^k \\ \Delta s^k \end{bmatrix}$$

⇒ In practice, we move in the direction for a certain stepsize  $\alpha^k > 0$

$$\text{Ex) } \alpha^k = \max_{\alpha > 0} \left\{ \begin{array}{l} x^h + \alpha \Delta x^k > 0 \text{ and} \\ s^h + \alpha \Delta s^k > 0 \end{array} \right\}$$

↳ The approach described above can produce iterates that get close to the boundary of the constraints ( $(x_i, y_i)$  primal-dual feasible and  $x_i s_i = 0$  for all  $i$ )

For this reason, researchers have developed path-following methods in which :

- a) The linear system is perturbed

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

where  $\sigma \in (0, 1)$  and  $\mu^k = \frac{(x^k)^T (s^k)}{m}$

- b)  $\alpha^k$  is computed under more restrictive conditions

Short-step path-following : Compute  $\alpha^k > 0$  so that  $x^{h+1} > 0$  and  $s^{h+1} > 0$

and  $\|X^{h+1} S^{h+1} e - \mu^{h+1} e\| \leq \gamma \mu^{h+1}$  for  $\gamma \in (0, 1)$

"The coefficients  $x_i^k s_i^k$  are decreasing in an uniform fashion"

Long-step path-following : Compute  $\alpha^k > 0$  so that  $x^{h+1} > 0$  and  $s^{h+1} > 0$

and  $x_i^{h+1} s_i^{h+1} \geq \gamma \mu^{h+1}$  for  $\gamma \in (0, 1)$

Every component of  $[x_i^{h+1} s_i^{h+1}]$  represents at least a fraction of  $\mu^{h+1}$

→ Avoids very small values of  $x_i^{h+1} s_i^{h+1}$  compared to the others

## 2) Barrier functions and IPIs

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

Idea: . Want to get to a solution from "the interior"

$$\{x^k\} \quad x^k > 0 \Rightarrow x^* \text{ solution}$$

. Approach: Replace the constraints  $x \geq 0$  by a barrier term in the objective to guarantee positivity

At every iteration  $k$ , we consider

$$(P_{\mu^k}) \underset{x \in \mathbb{R}^m}{\text{minimize}} \quad c^T x - \mu^k \sum_{i=1}^m \ln(x_i) \quad \text{s.t. } Ax = b \\ x \geq 0$$

for some  $\mu^k > 0$

$\hookrightarrow x \mapsto c^T x - \mu^k \sum_{i=1}^m \ln(x_i)$  is nonlinear

equal to  $+\infty$  at every  $x$  with a zero coordinate  
and it excludes de facto negative coordinates

Barrier IPIs compute a sequence of solutions of subproblems  $(P_{\mu^k})$  for decreasing values of  $\mu^k$

As  $\mu^k \rightarrow 0$ , the solution gets closer to that of the original problem (but it remains in the interior!)

$\hookrightarrow$  there exist primal, dual, and primal-dual barrier methods depending on which problem(s) are solved