Zewail City of Science and Technology University of Science and Technology Math 404 - fall 2019

# Report 3

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# 1 Approach

To implement interior point method along with Mehrotra, we need to implement three following algorithm from Numerical Optimization by Jorge Nocedal, Steve Wright chapter 14. I am still working on code and you will find updates of it on this google drive link

## 2| First Algorithm: Primal-Dual Path-Following

Framework 14.1 (Primal-Dual Path-Following).

**Given** 
$$(x^0, \lambda^0, s^0)$$
 with  $(x^0, s^0) > 0$ ;

for 
$$k = 0, 1, 2, ...$$

Choose  $\sigma_k \in [0, 1]$  and solve

$$\begin{bmatrix} 0 & A^T & I \\ A & 0 & 0 \\ S^k & 0 & X^k \end{bmatrix} \begin{bmatrix} \Delta x^k \\ \Delta \lambda^k \\ \Delta s^k \end{bmatrix} = \begin{bmatrix} -r_c^k \\ -r_b^k \\ -X^k S^k e + \sigma_k \mu_k e \end{bmatrix},$$

where 
$$\mu_k = (x^k)^T s^k / n$$
;

Set

$$(x^{k+1}, \lambda^{k+1}, s^{k+1}) = (x^k, \lambda^k, s^k) + \alpha_k(\Delta x^k, \Delta \lambda^k, \Delta s^k),$$

choosing  $\alpha_k$  so that  $(x^{k+1}, s^{k+1}) > 0$ .

end (for).

# 2 | Second Algorithm: Long-Step Path-Following

#### CENTRAL PATH NEIGHBORHOODS AND PATH-FOLLOWING METHODS

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Algorithm 14.2 (Long-Step Path-Following). Given \gamma, \sigma_{\min}, \sigma_{\max} with \gamma \in (0, 1), 0 < \sigma_{\min} \le \sigma_{\max} < 1, and (x^0, \lambda^0, s^0) \in \mathcal{N}_{-\infty}(\gamma); for k = 0, 1, 2, \ldots Choose \sigma_k \in [\sigma_{\min}, \sigma_{\max}]; Solve (14.10) to obtain (\Delta x^k, \Delta \lambda^k, \Delta s^k); Choose \alpha_k as the largest value of \alpha in [0, 1] such that
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$$(x^k(\alpha), \lambda^k(\alpha), s^k(\alpha)) \in \mathcal{N}_{-\infty}(\gamma);$$
 Set  $(x^{k+1}, \lambda^{k+1}, s^{k+1}) = (x^k(\alpha_k), \lambda^k(\alpha_k), s^k(\alpha_k));$  end (for).

### 3 Third Algorithm: Predictor – Corrector (Mehrotra)

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Algorithm 14.3 (Predictor-Corrector Algorithm (Mehrotra [207])). Calculate (x^0, \lambda^0, s^0) as described above;  \begin{aligned} &\text{for } k = 0, 1, 2, \dots \\ &\text{Set } (x, \lambda, s) = (x^k, \lambda^k, s^k) \text{ and solve } (14.30) \text{ for } (\Delta x^{\text{aff}}, \Delta \lambda^{\text{aff}}, \Delta s^{\text{aff}}); \\ &\text{Calculate } \alpha_{\text{aff}}^{\text{pri}}, \alpha_{\text{aff}}^{\text{dual}}, \text{ and } \mu_{\text{aff}} \text{ as in } (14.32) \text{ and } (14.33); \\ &\text{Set centering parameter to } \sigma = (\mu_{\text{aff}}/\mu)^3; \\ &\text{Solve } (14.35) \text{ for } (\Delta x, \Delta \lambda, \Delta s); \\ &\text{Calculate } \alpha_k^{\text{pri}} \text{ and } \alpha_k^{\text{dual}} \text{ from } (14.38); \\ &\text{Set} \end{aligned}  Set  x^{k+1} = x^k + \alpha_k^{\text{pri}} \Delta x, \\ &(\lambda^{k+1}, s^{k+1}) = (\lambda^k, s^k) + \alpha_k^{\text{dual}} (\Delta \lambda, \Delta s); \end{aligned}  end (for).
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Figure 1 Fibonacci testing