

11.3 BARRIER METHOD (PATH-FOLLOWING METHOD)

SOLVING LARGE t DIFFICULT SO:

BARRIER METHOD ALGORITHM

INPUT x STRICTLY FEASIBLE (SUBOPT)

$t > 0$

$\mu > 0$ INCREASE OF t

$\epsilon > 0$ TOLERANCE

CONTAINS:

$x^*(t)$ MINIMIZER OF $t f(x) + \phi(x)$

→ ITERATIVE METHOD WITH INITIAL POINT x

UPDATE: $x := x^*(t)$

STOP? : IF $m/t < \epsilon$ STOP

$t := \mu t$

OUTER ITERATION : ONCE CYCLE

INNER ITERATION -- CYCLE INSIDE

CENTERING STEP - USE NEWTON METHOD

NOTES;

MUST BE THE INNER LOOP EXACT?

- NO, BUT THEN MAKE DUAL SOLUTION. BUT STILL CONVERGES
- BUT FROM CLOSE SOL. TO "EXACT", NEWTON IS FAST

HOW TO PICK m ?

LARGE m - FEW OUTER ITERATIONS

SMALL m - FEW INNER ITERATIONS

(TYPICAL $m = 10 - 20$ WITH NEWTON)

HOW TO PICK ϵ FOR INPUT?

LARGE ϵ - DIFFICULT FIRST STEP

SMALL ϵ - MANY OUTER ITERATIONS

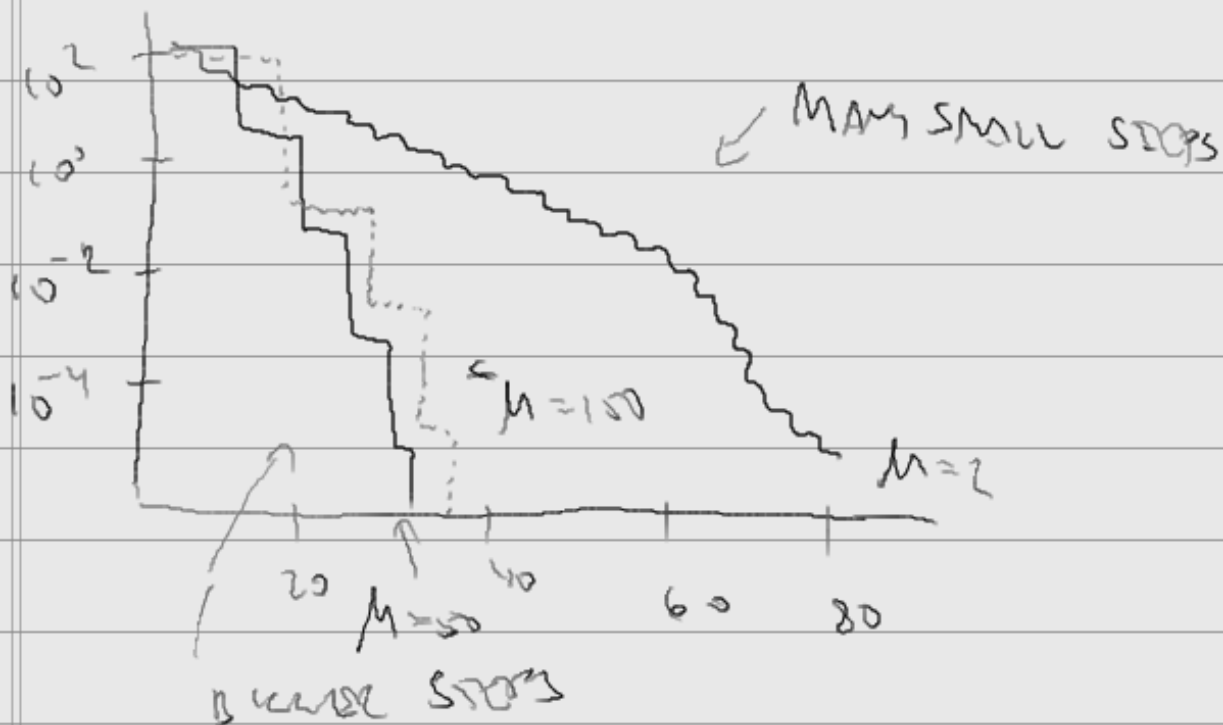
SET $\epsilon \approx m/\epsilon \approx f(x) - f(x^*)$

OR IF λ DUAL FEASIBLE

IF ROW

USE $\epsilon = m / (f(x) - h(\lambda))$

EXAMPLE:



DETAILS IN BOOK IN SECTION 11.3.2

CONVERGENCE ANALYSIS:

OF OUTER ITERATIONS:

$$\text{GAP: } \frac{m}{M^k t} < \epsilon \quad \leadsto \quad \left\lceil \frac{\log\left(\frac{m}{\epsilon t}\right)}{\log M} \right\rceil = k$$

THE FIRST
Iter.

INNER ITERATIONS BASED ON NEWTON'S METHOD

NEWTON STEP FOR KKT

1 CENTERING STEP: $[H\lambda(x_k)](x_{k+1}-x_k) = -\sigma\lambda(x_k)$

$$(\epsilon H\lambda(x_k) + H\phi(x_k))(x_{k+1}-x_k) = -\epsilon\sigma\lambda(x_k) - \sigma\phi(x_k)$$

SOLVING KKT:

$$\nabla\lambda(x) + \sum \lambda_i \nabla g_i(x) = 0$$

$$-\lambda_i \nabla g_i(x) = 1/\epsilon$$

$$\lambda_i \geq 0$$

$$\rightarrow \text{REPLACE } \lambda_i = \frac{1}{-\epsilon \nabla g_i(x)}$$

$$\nabla\lambda(x) + \sum \frac{1}{-\epsilon \nabla g_i(x)} \nabla g_i(x) = 0$$

$$f(x_k + \nu) \approx$$

$$\approx f(x_k) + \nu \nabla f(x_k)$$

↑

IN NEWTON METHOD FROM x_k USE APPROX

$$\nabla\lambda(x+\nu) + \sum_{i=1}^m \frac{1}{-\epsilon \nabla g_i(x+\nu)} \nabla g_i(x+\nu) \approx$$

$$\approx \nabla\lambda(x_k) + \sum \frac{1}{-\epsilon \nabla g_i(x_k)} \nabla g_i(x_k) + H\lambda(x_k) \nu$$

$$+ \sum \frac{1}{-\epsilon \nabla g_i(x_k)} H g_i(x_k) \nu + \sum \frac{1}{-\epsilon \nabla g_i(x_k)} \nabla g_i(x_k) \nabla g_i(x_k)^T \nu$$

= 0

$$H\nu = -y$$

$$\nabla_{x_i} H = H_{\Delta}(x_k) + \frac{1}{\epsilon} H\phi(x_k)$$

NOTE:

$$\nabla\phi(x) = \sum \frac{1}{-g_i(x)} \nabla g_i(x)$$

$$H\phi(x) = \sum \frac{1}{g_i(x)^2} \nabla g_i(x) \nabla g_i(x)^T + \nabla^2\phi(x) + \sum \frac{1}{-g_i(x)} H_{g_i(x)}$$

$$\left(H_{\Delta}(x_k) + \frac{1}{\epsilon} H\phi(x_k) \right) \nu = \nabla\phi(x_k) + \frac{1}{\epsilon} \nabla\phi(x_k)$$

IF $\nu = (x_{k+1} - x_k)$, GIVES

EXACTLY THE SAME THING

⇒ ITERATION OF CENTERING STEP IS
SAME AS ITERATION OF NEWTONS
METHOD ON KKT.

11.4 FEASIBILITY & PHASE I METHODS

STARTING POINT REQUIRED FOR INTERPOINT.

PHASE I - FIND INTERIOR POINT x

PHASE II - USE x AS STARTING POINT

BASIC METHOD:

(V)

$$\begin{array}{ll} \text{MIN } s & s \in \mathbb{R} \\ \text{ST. } g_i(x) \leq s \end{array}$$

IF SOLUTION (x^*, s^*) WHERE $s^* < 0$,
 x^* IS STRICTLY FEASIBLE.

$$s^* > 0,$$

NOT FEASIBLE AT ALL.

$$s^* = 0,$$

FEASIBLE BUT NOT STRICTLY

(VI)

$$\text{MIN } \sum s$$

$$\text{ST. } g_i(x) \leq s_i$$

$$s_i \geq 0$$

\rightarrow FOR INFEASIBLE PROBLEM SATISFYABLE

\rightarrow THAN (VI)

BY ADDING TO (V)

MIN s

$$\text{ST. } g_i(x) \leq s \quad \dots \quad g_i(x) - s \leq 0$$

$$f(x) \leq M \quad \dots \quad f(x) - M \leq 0$$

WHERE $M > f(x^0)$

\Rightarrow CENTRAL PATHS OF AND (P) INTERSECT

MIN $ts + \phi(x, s)$

CHARACTERIZATION:

$$\left[t f(x^0(t)) + \sum_{i=1}^m \frac{1}{s - g_i(x^0(t))} \nabla g_i(x^0(t)) \right]$$

$$x: \frac{1}{M - f_0(x)} \nabla f_0(x) + \sum_{i=1}^m \frac{1}{s - g_i(x)} \nabla g_i(x) = 0$$

$$s: t - \sum_{i=1}^m \frac{1}{s - g_i(x)} = 0$$

FOR $s = 0$ AND $t = \frac{1}{M - f_0(x)}$ WE

GET POINT ON CENTRAL PATH FOR (P).