

NEWTON TYPE ALGORITHMS IN OPTIMIZATION

2c

EQUALITY CONSTRAINED OPTIMIZATION

Equality constraints only, first

$$\left. \begin{array}{l} f: \mathbb{R}^n \rightarrow \mathbb{R} \\ g: \mathbb{R}^n \rightarrow \mathbb{R}^m \end{array} \right\} \begin{array}{l} \text{(Both assumed to} \\ \text{be smooth)} \end{array}$$

$$\left. \begin{array}{l} \min_{w \in \mathbb{R}^n} f(w) \\ \text{s.t. } g(w) = 0 \end{array} \right\}$$

WE CAN USE NEWTON'S TYPE METHODS, OR SOME VARIANT OF IT, TO SOLVE THE KKT CONDITIONS (which are nonlinear)

$$\left. \begin{array}{l} \nabla_w \mathcal{L}(w, \lambda) = 0 \\ g(w) = 0 \end{array} \right\} \begin{array}{l} \text{from } \mathcal{L}(w, \lambda, \mu) = f(w) - \lambda^T g(w) - \mu^T h(w) \\ \text{(THE LAGRANGIAN FUNCTION)} \end{array} = 0$$

THIS IS NOW A ROOT-FINDING PROBLEM

WE SIMPLIFY THE NOTATION BY DEFINING THE FOLLOWING TERMS

$$z = \begin{bmatrix} w \\ \lambda \end{bmatrix} \quad \text{AND} \quad R(z) = \begin{bmatrix} \nabla_w \mathcal{L}(w, \lambda) \\ g(w) \end{bmatrix},$$

$$\text{SO THAT } z \in \mathbb{R}^{N+N_1} \quad \text{AND} \quad R: \mathbb{R}^{N+N_1} \longrightarrow \mathbb{R}^{N+N_2}$$

WE CAN WRITE THE ROOT-FINDING PROBLEM COMPACTLY AS

$$R(z) = 0$$

CONSIDER AN INITIAL SOLUTION z_0 , THE NEWTON'S METHOD GENERATES A SEQUENCE OF ITERATES $(z_k)_{k=0}^\infty$ BY LINEARIZING THE NONLINEAR EQUATION AT THE CURRENT ITERATION POINT

$$R(z_k) + \frac{\partial R}{\partial z} \Big|_{z_k} (z - z_k) = 0$$

AND THE NEXT ITERATE IS THE SOLUTION TO THE LINEARISED EQ.

$$z_{k+1} = z_k - \left[\frac{\partial R}{\partial z} \Big|_{z_k} \right]^{-1} R(z_k)$$

IN TERMS OF GRADIENTS

$$\nabla_w \mathcal{L}(w_k, \lambda_k) + \underbrace{\nabla_w^2 \mathcal{L}(w, \lambda)(w - w_k)}_{\text{LINEARISATION WRT } w} - \underbrace{\nabla_g(w_k)(\lambda - \lambda_k)}_{\text{LINEARISATION WRT } \lambda} = 0$$

By definition :

$$\begin{cases} \lambda_{k+1} = \lambda_k + \Delta \lambda_k \\ \nabla \mathcal{L}(w_k, \lambda_k) = \nabla f(w_k) - \nabla g(w_k) \end{cases}$$

To calculate $\nabla \lambda_k$ and ∇w_k is equivalent to solving

$$\begin{bmatrix} \nabla f(w_k) \\ g(w_k) \end{bmatrix} + \begin{bmatrix} \nabla^2 \mathcal{L} & \nabla g \\ \nabla g^\top & 0 \end{bmatrix} \begin{bmatrix} \Delta w_k \\ -\lambda_{k+1} \end{bmatrix} = 0$$

FOR EQUALITY CONSTRAINED OPTIMIZATION PROBLEMS, THE LINEAR SYSTEM $\nabla R(z_k) + \frac{\partial R}{\partial z} \Big|_{z_k} (z - z_k) = 0$ HAS A SPECIFIC FORM

RESIDUALS

MATRIX OF SECOND DERIVATIVES

$$\begin{bmatrix} \nabla_w \mathcal{L}(w_k, x_k) \\ g(w_k) \end{bmatrix}^+ \underbrace{\begin{bmatrix} \nabla_w^2 \mathcal{L}(w_k, x_k) & \nabla g(w_k) \\ \nabla g(w_k)^T & 0 \end{bmatrix}}_{\text{KKT MATRIX}} \begin{bmatrix} w - w_k \\ x - x_k \end{bmatrix} = 0$$

BY USING THE DEFINITION $\nabla_w \mathcal{L}(w_k, x_k) = \nabla f(w_k) + \nabla g(w_k) \lambda_k$

WE CAN REWRITE THE SYSTEM AS

$$\begin{bmatrix} \nabla f(w_k) \\ g(w_k) \end{bmatrix}^+ \underbrace{\begin{bmatrix} \nabla_w^2 \mathcal{L}(w_k, x_k) & \nabla g(w_k) \\ \nabla g(w_k)^T & 0 \end{bmatrix}}_{\text{KKT MATRIX}} \begin{bmatrix} w - w_k \\ x \end{bmatrix} = 0$$

There is no dependence on λ_k

NEWTON-TYPE
vs

EXACT NEWTON METHODS

The only dependence on λ_k is via the HESSIAN matrix (WHICH CAN BE APPROXIMATED)

let $B_k = \nabla_w^2 \mathcal{L}(w_k, x_k)$, we have the NEWTON-TYPE ITERATION

$$\begin{bmatrix} w_{k+1} \\ \lambda_{k+1} \end{bmatrix} = \begin{bmatrix} w_k \\ 0 \end{bmatrix} - \begin{bmatrix} B_k & \nabla g(w_k) \\ \nabla g(w_k)^T & 0 \end{bmatrix}^{-1} \begin{bmatrix} \nabla f(w_k) \\ g(w_k) \end{bmatrix}$$

Invertibility is required

(LICQ) + (SECOND ORDER OPTIMALITY CONDITION)
+ (STRICT COMPLEMENTARITY)

SUFFICIENT

No 12

INEQUALITY CONSTRAINED OPTIMIZATION

minimise $f(w)$
 $w \in \mathbb{R}^N$
 subject to $g(w) = 0$
 $h(w) \geq 0$

$$\left\{ \begin{array}{l} f: \mathbb{R}^N \rightarrow \mathbb{R} \\ g: \mathbb{R}^N \rightarrow \mathbb{R}^{N_g} \\ h: \mathbb{R}^N \rightarrow \mathbb{R}^{N_h} \end{array} \right.$$

All assumed to be smooth

TO SOLVE A GENERAL NONLINEAR OPTIMISATION PROBLEM WITH INEQUALITY CONSTRAINTS THERE EXIST TWO MAIN APPROACHES

- (NONLINEAR) INTERIOR POINT METHODS
- SEQUENTIAL QUADRATIC PROGRAMMING

BOTH APPROACHES AIM AT SOLVING THE KKT CONDITIONS

$$\begin{aligned} \nabla f(w^*) + \nabla g(w^*) \lambda^* + \nabla h(w^*) \mu^* &= 0 \\ g(w^*) &= 0 \\ h(w^*) &\leq 0 \\ \mu^* &\geq 0 \\ \underbrace{\mu^* h_i(w^*)}_{\geq 0} &= 0 \quad (i=1, 2, \dots, N_h) \end{aligned}$$

NONLINEAR INTERIOR POINT METHODS REPLACE THE KINKINESS OF THE NON-SMOOTH COMPLEMENTARITY CONDITIONS WITH A SMOOTH APPROXIMATION

→ THE L-SHAPED SET IS REPLACED BY A HYPERBOLA
 (typically)

$$\begin{aligned} \text{TO GET } \nabla f(w^*) + \nabla g(w^*) \lambda^* + \nabla h(w^*) \mu^* &= 0 \\ g(w^*) &= 0 \\ \underbrace{\mu^* h_i(w^*) + \gamma}_{\geq 0} &= 0 \end{aligned}$$

SO THAT $-h_i(\mu^*)$ AND μ^* ARE BOTH POSITIVE
 AND ON AN HYPERBOLA FOR γ SMALL

SEQUENTIAL QUADRATIC PROGRAMMING AT EACH ITERATION IT SOLVES AN INEQUALITY CONSTRAINED QUADRATIC PROGRAM OBTAINED BY LINEARISING THE OBJECTIVE FUNCTION AND THE CONSTRAINT FUNCTIONS

$$\text{minimise}_{w \in \mathbb{R}^n} \quad \nabla f(w)(w - w_k) + \frac{1}{2}(w - w_k)^T B(w - w_k)$$

$$\text{subject to} \quad g(w_k) + \nabla g^T(w_k)(w - w_k) = 0 \\ h(w_k) + \nabla h^T(w_k)(w - w_k) \leq 0$$

