

## NEWTON TYPE ALGORITHMS IN OPTIMIZATION

### EQUALITY CONSTRAINED OPTIMIZATION

Equality constraints only, first

$$f: \mathbb{R}^n \rightarrow \mathbb{R}$$

$$g: \mathbb{R}^n \rightarrow \mathbb{R}^m \quad (\text{Both assumed to be smooth})$$

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

2c

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

$$\begin{cases} \nabla_w \mathcal{L}(w, \lambda) = 0 \\ g(w) = 0 \end{cases} ; \text{ from } \mathcal{L}(w, \lambda, \mu) = f(w) - \lambda^T g(w) - \mu^T h(w) \quad (= 0) \\ (\text{THE LAGRANGIAN FUNCTION}) \\ \text{THIS IS NOW A ROOT-FINDING PROBLEM}$$

WE SIMPLIFY THE NOTATION BY DEFINING THE FOLLOWING TERMS

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

$$\text{SO THAT } z \in \mathbb{R}^{N+N_\lambda} \quad \text{AND} \quad R: \mathbb{R}^{N+N_\lambda} \rightarrow \mathbb{R}^{N+N_\lambda}$$

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 LINEARIZED EQ.

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

NOTE

IN TERMS OF GRADIENTS

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

By definition :

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

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

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

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

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

RESIDUALS → MATRIX OF SECOND DERIVATIVES

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

WE CAN REWRITE THE SYSTEM AS

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

There is no dependence on  $\lambda_k$

NEWTON-TYPE  
vs

EXACT NEWTON METHODS

The only dependence on  $\lambda_k$   
is via the HESSIAN matrix  
(WHICH CAN BE APPROXIMATED)

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

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

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

Note

## 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

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

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

SO THAT  $-h_i(w^*)$  AND  $\mu_i^*$  ARE BOTH POSITIVE  
AND ON AN HYPERBOLA FOR  $\gamma_i$  SMALL

No 13

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

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

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

$$h(w_k) + \nabla h(w_k)^T (w - w_k) \leq 0$$