

NUMERICAL OPTIMISATION

ROOT FINDING METHODS

- Generalities
- Convergence rates, contraction, invariance; conditions for convergence

NONLINEAR OPTIMISATION

- Classes of problems
- First order optimality conditions
- Second order optimality conditions

2

NEWTON-TYPE ALGORITHMS

- Equality constrained problems, convergence
- Inequality constrained problems
- Globalisation techniques

DERIVATIVES

- Algorithm differentiation
- Forward mode
- Backward mode

PARAMETER ESTIMATION

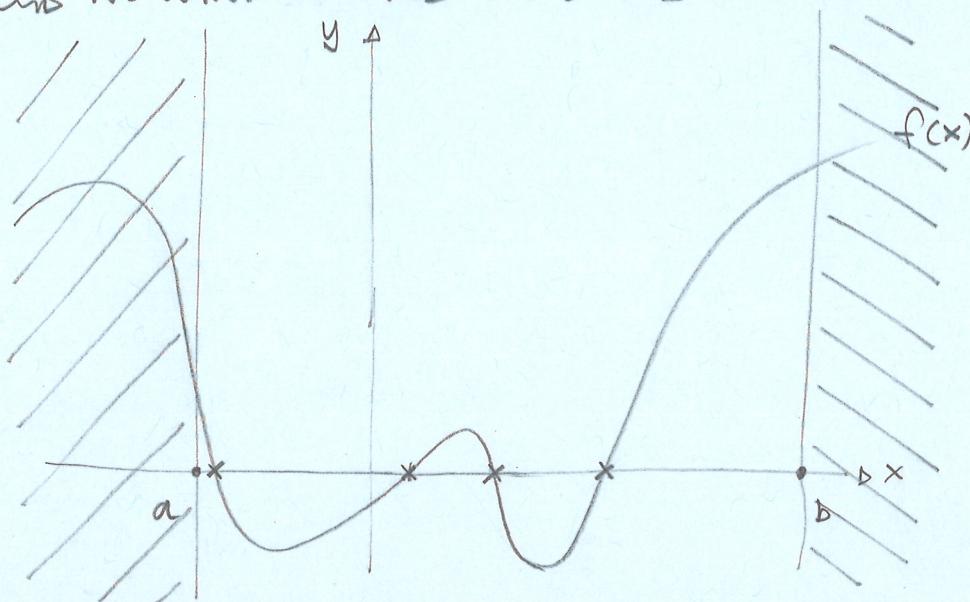
- Least-squares
- Convex penalty terms

ROOT FINDING AND NEWTON-TYPE METHODS

WE CONSIDER THE PROBLEM OF FINDING THE ZEROS OF SOME FUNCTION

$$f: [a, b] \subset \mathbb{R} \rightarrow \mathbb{R}$$

AND WE WANT TO FIND $x \in [a, b]$ such that $f(x) = 0$

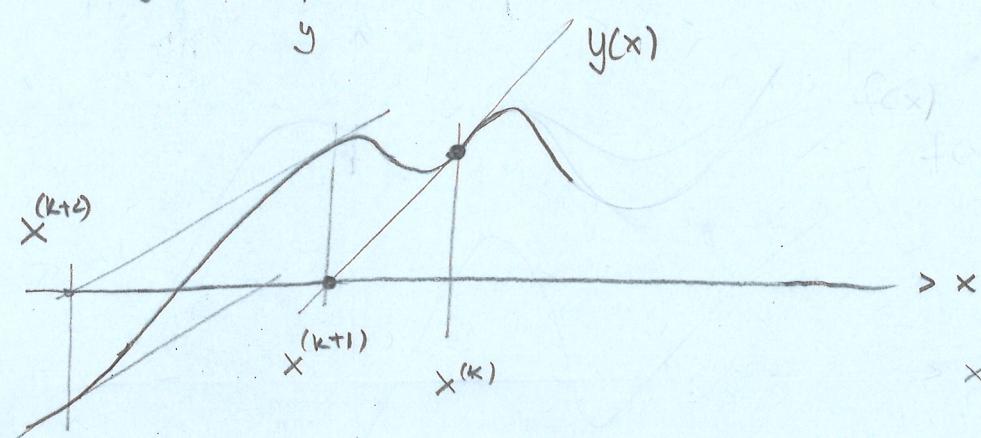


2a



WE KNOW THE EQUATION OF THE TANGENT LINE TO THE FUNCTION AT POINT $x^{(k)}$

$$y(x) = f[x^{(k)}] + f'[x^{(k)}][x - x^{(k)}]$$



- WE CAN FIND THE POINT WHERE THE TANGENT LINE EQUALS ZERO

- AND ITERATE

- THE SEQUENCE $\{x^{(k)}\}$ IS THE NEWTON'S METHOD FOR FINDING THE ZEROS OF FUNCTION f

$$\begin{aligned} 0 &= f[x^{(k)}] + f'[x^{(k)}][x - x^{(k)}] \\ &= f[x^{(k)}] + f'[x^{(k)}]x - f'[x^{(k)}]x^{(k)} \\ \Rightarrow x &= \frac{-f[x^{(k)}] + f'[x^{(k)}]x^{(k)}}{f'[x^{(k)}]} \\ &= x^{(k)} - f(x^{(k)})/f'(x^{(k)}) \end{aligned}$$

RF0

NOW CONSIDER A SET OF FUNCTIONS, NOW ALSO TWIVARIATE

$$\left\{ \begin{array}{l} f_1(x_1, x_2, \dots, x_n) = 0 \\ f_2(x_1, x_2, \dots, x_n) = 0 \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) = 0 \end{array} \right.$$

WE WANT TO FIND THE VALUE $x \in \mathbb{R}^n$ SUCH THAT $f(x) = 0$

\Rightarrow WE CAN EXPEND NEWTON'S METHOD

\Rightarrow WE REPLACE THE FIRST DERIVATIVE OF f WITH THE JACOBIAN OF f

$$\text{let } J_f = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \cdots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

$$\text{WE HAD } x^{(k+1)} = x^{(k)} - \frac{f[x^{(k)}]}{f'[x^{(k)}]}$$

$$\text{IT BECOMES } x^{(k+1)} = x^{(k)} - J^{-1}[x^{(k)}] f[x^{(k)}] \quad (**)$$

Example

$$\left\{ \begin{array}{l} f_1(x_1, x_2) = x_1^2 + x_2^2 - 1 \\ f_2(x_1, x_2) = \sin(\pi/2 x_1) + x_2^3 = 0 \end{array} \right. \quad J = ?$$

The system has two solutions $\left\{ \begin{array}{l} (0.47, -0.88) \\ (-0.47, 0.88) \end{array} \right.$

WHY IS A METHOD FOR ROOT FINDING IMPORTANT IN OPTIMISATION?

AND WE CAN HAVE $f(x) = 0$ WITH $\nabla f(x)$ THE GRADIENT OF
SOME OBJECTIVE FUNCTION $f(x)$

AND THE JACOBIAN OF THE SYSTEM IS THIS GIVEN BY THE
HESSEIAN $H(x)$ OF $f(x)$

WE GET THE ITERATES $x^{(k+1)} = x^{(k)} - H^{-1}[x^{(k)}] \nabla f[x^{(k)}]$

Example

$$R(z) = z^{16} - 1 \quad \text{such that } \frac{\partial R(z)}{\partial z} = 16z^{15}$$

→ THE ITERATES OF THE NEWTON'S METHOD ARE

$$z_{k+1} = z_k - \underbrace{(16z_k^{15})^{-1}}_{\text{Jacobian}} \underbrace{(z_k^{16} - 1)}_{R(z)}$$

→ ALTERNATIVELY, WE COULD USE AN APPROXIMATION OF THE JACOBIAN FOR EXACTLY A CONSTANT VALUE, $M_k = 16$

ROOT FINDING AND NEWTON'S TYPE METHODS

POTENTIALLY NONLINEAR

WE ARE INTERESTED IN SOLVING A SYSTEM OF EQUATIONS $R(\mathbf{z}) = \mathbf{0}$
AND FUNCTION $R : \mathbb{R}^n \rightarrow \mathbb{R}^n$ (so $\mathbf{z} \in \mathbb{R}^n$)

WE KNOW THAT THIS FUNCTION IS DIFFERENTIABLE

Example

- i) $R(\mathbf{z}) = \mathbf{z}^6 - \text{const}$ (1 variable, 1 equation) $\mathbf{z} \in \mathbb{R}$
 ii) $R(\mathbf{z}) = \begin{pmatrix} \mathbf{z}_1^n - \mathbf{z}_2 \\ \mathbf{z}_2^n - \text{const} \end{pmatrix} \begin{matrix} R_1(\mathbf{z}) \\ R_2(\mathbf{z}) \end{matrix}$ (2 variables, 2 equations) $\mathbf{z} \in \mathbb{R}^2$

The derivative(s) of $R(\mathbf{z})$

$$\text{i)} \frac{\partial R}{\partial \mathbf{z}} = 16\mathbf{z}^{15}$$

$$\text{ii)} \frac{\partial R}{\partial \mathbf{z}} = \begin{bmatrix} 4\mathbf{z}_1^3 & -1 \\ 0 & 4\mathbf{z}_2^3 \end{bmatrix} \quad \begin{array}{l} \frac{\partial R_1}{\partial z_1} \\ \frac{\partial R_1}{\partial z_2} \\ \hline \frac{\partial R_2}{\partial z_1} \\ \frac{\partial R_2}{\partial z_2} \end{array} \quad \begin{array}{l} \frac{\partial R_1}{\partial z_1} \\ \frac{\partial R_1}{\partial z_2} \\ \hline \frac{\partial R_2}{\partial z_1} \\ \frac{\partial R_2}{\partial z_2} \end{array} \quad \text{The Jacobian matrix}$$

IDEA BEHIND NEWTON'S METHODS \rightsquigarrow USE A TAYLOR SERIES (1st)
EXPANSION OF R AND SOLVE

1. GUESS $\mathbf{z}_k \in \mathbb{R}^n$
2. GET NEW \mathbf{z}_{k+1} AS SOLUTION OF $R(\mathbf{z}_k) + \underbrace{\frac{\partial R}{\partial \mathbf{z}}(\mathbf{z}_k)(\mathbf{z} - \mathbf{z}_k)}_{\approx R(\mathbf{z})} \approx R(\mathbf{z}) = 0$

* THIS CAN BE EXPLICITLY COMPUTED

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

1st ORDER TAYLOR
SERIES APPROX OF $R(\mathbf{z})$

NEWTON'S ITERATION

THE JACOBIAN MUST BE INVERTIBLE

J_k

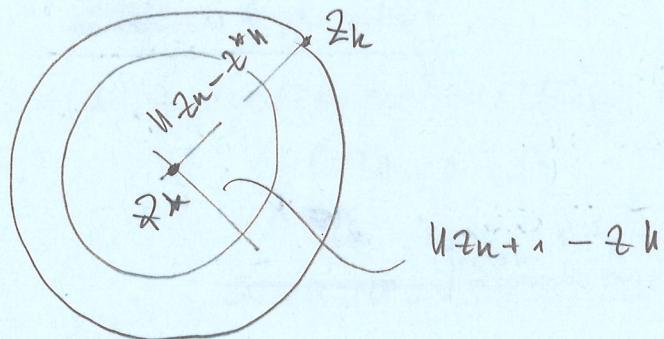
3. REPEAT 2.

\hookrightarrow STOPS IF $J_k = 0$ OR $R(\mathbf{z}_k) = 0$

RF1

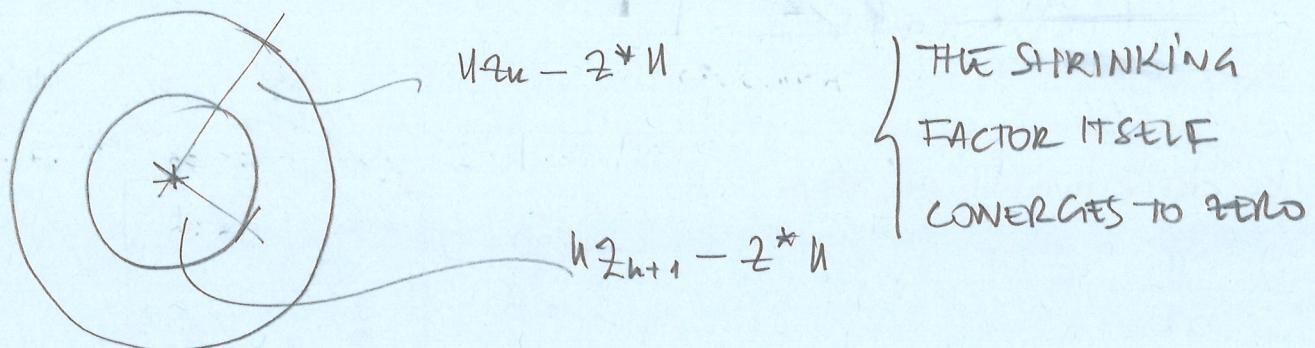
Q-LINEAR CONVERGENCE

(After local convergence rate, when in class)

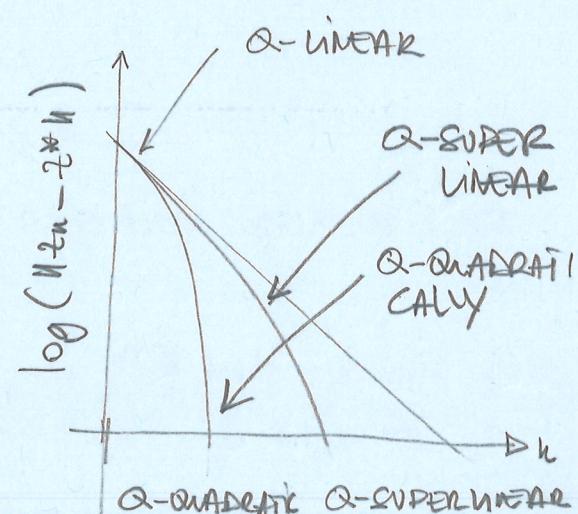
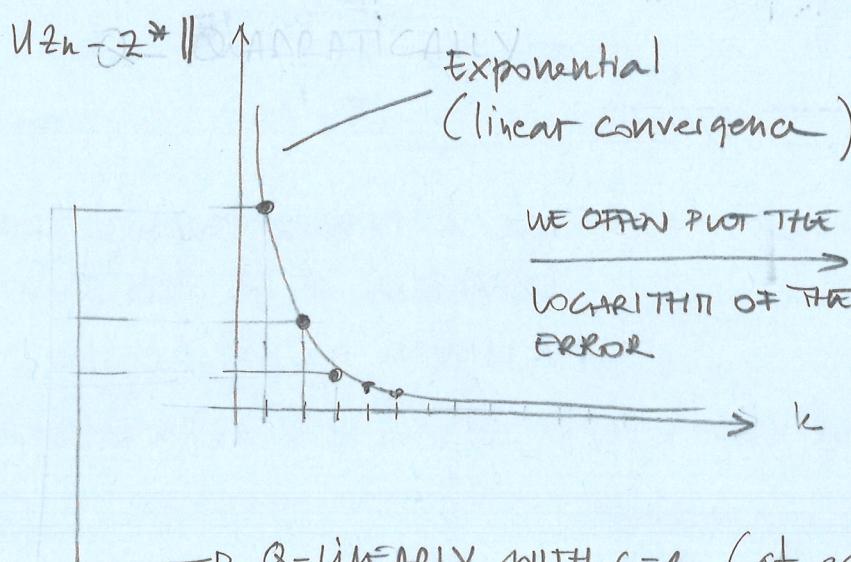


{ THE NEXT ITERATE MUST
BE WITHIN THE k-TH
BALL (SHRINKS)

Q-SUPERLINEAR CONVERGENCE



{ THE SHRINKING
FACTOR ITSELF
CONVERGES TO ZERO



→ Q-LINEARLY WITH $c=2$ (at each iteration we contract by a factor equal $c=2$)
(it is exponential, so linear in log scale)

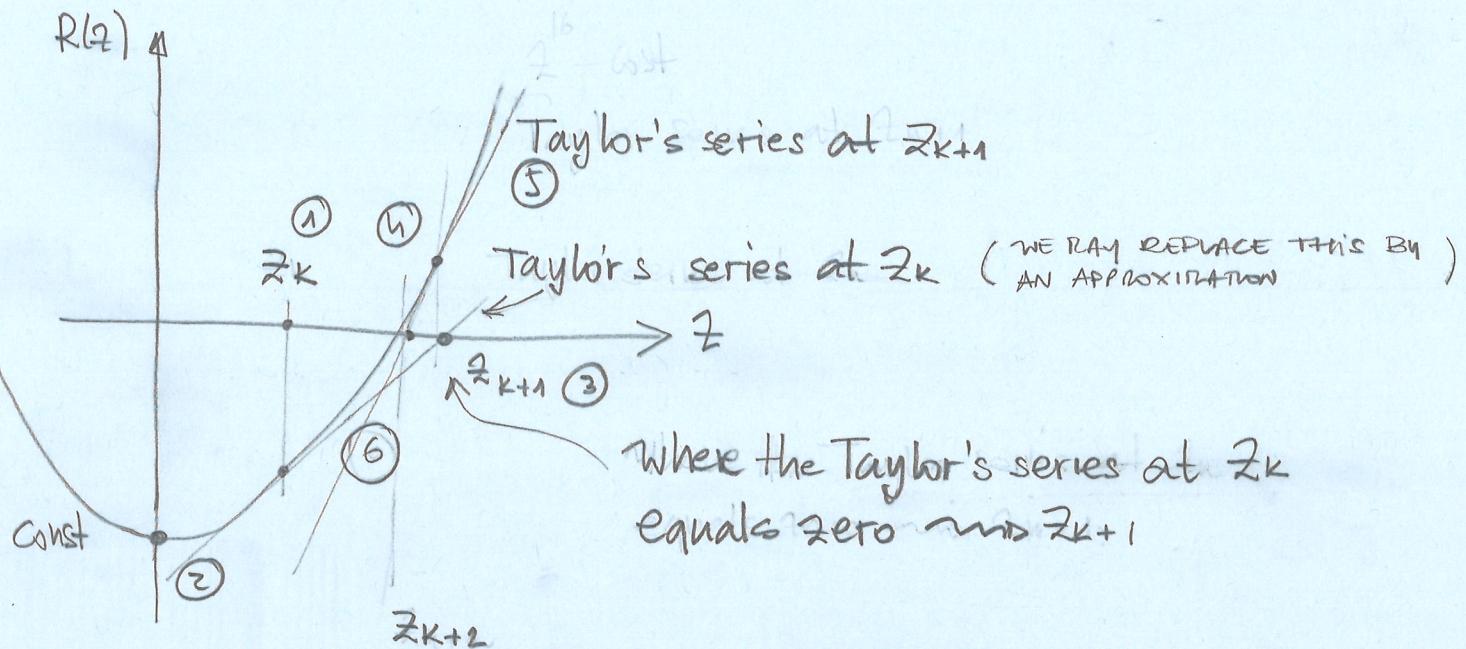
On a general level, we can use some invertible approximation M_k of the Jacobian J_k

$$\rightsquigarrow z_{k+1} = z_k - [M_k]^{-1} R(z_k)$$

APPROXIMATES THE JACOBIAN

NEWTON-TYPE ITERATION

GENERAL FORM OF A NEWTON'S TYPE METHOD $(M_k \approx \frac{\partial R}{\partial z} \Big|_{z_k})$

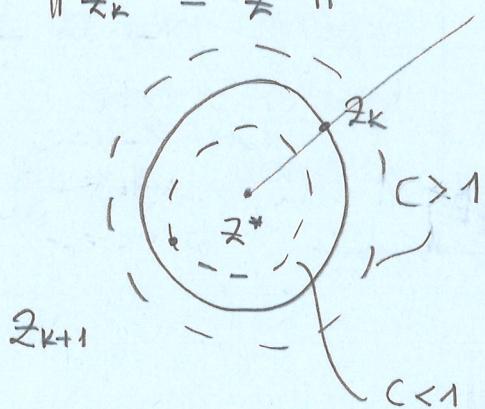


THE SEQUENCE OF ITERATES $\{z_k\}_{k=0}^{\infty}$ WILL CONVERGE TO THE SOLUTION

- \rightsquigarrow WE NEED TO DEFINE THE CONVERGENCE RATE!
- \rightsquigarrow (CONDITIONS FOR CONVERGENCE / DIVERGENCE)

$$\frac{\|z_{k+1} - z^*\|}{\|z_k - z^*\|} \leq c$$

α -LINEAR CONVERGENCE



C IS A SHRINKING FACTOR
AND IT IS OF FIXED SIZE

α -SUPERLINEAR CONVERGENCE

$$\frac{\|z_{k+1} - z^*\|}{\|z_k - z^*\|} \leq c_k$$



c_k IS A SHRINKING FACTOR
AND ITSELF SHRINKS (VARIABLE SIZE)
TO ZERO (IT GETS INFINITELY SMALLER)

$$\|z_{k+1} - z^*\| \leq$$

α -QUADRATIC CONVERGENCE

$$c \|z_k - z^*\| \|z_k - z^*\|$$

$$\underbrace{\quad}_{\|z_k - z^*\|^2}$$

$$\|z_k - z^*\|^2$$

LOCAL CONVERGENCE RATES

CONVERGES TO

Def. (Types of convergence rates) Assume $z_k \in \mathbb{R}^N$ and $z_k \xrightarrow{k \rightarrow \infty} z^*$ for $k \rightarrow \infty$

We say:

① (z_n) is convergent to z^* Q-LINEARLY

IFF $\exists C < \infty$ such that $\|z_{n+1} - z^*\| \leq C \|z_n - z^*\|$

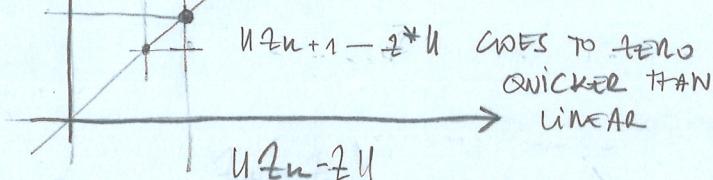
(for all $k \geq k_0$)

CONVERGE TO
SOMETHING FOR
 $k \rightarrow \infty$

WITH $C < 1$

$$\Rightarrow \limsup_{k \rightarrow \infty} \frac{\|z_{k+1} - z^*\|}{\|z_k - z^*\|} < 1$$

"QUOTIENT"



② (z_k) is convergent to z^* Q-SUPERLINEARLY (STRONGER)

IFF $\exists c_k$ such that $\|z_{n+1} - z^*\| \leq c_k \|z_n - z^*\|$ WITH $c_k \rightarrow 0$

↓ THE SEQUENCE $\{c_k\}$ CONVERGE TO ZERO

$$\Rightarrow \limsup_{k \rightarrow \infty} \frac{\|z_{n+1} - z^*\|}{\|z_n - z^*\|} = 0 \quad (\text{for all } k \geq k_0)$$

③ (z_n) is convergent to z^* Q-QUADRATICALLY

IFF $\exists C < \infty$ such that $\|z_{k+1} - z^*\| \leq C \|z_k - z^*\|^2$

(for all $k \geq k_0$)

$$\leq C \|z_k - z^*\| \|z_k - z^*\|$$

THE SIZE OF THE
SHRINKING DEPENDS ON THE
CLOSENESS (LINEARLY)

$$\Rightarrow \limsup_{k \rightarrow \infty} \frac{\|z_{k+1} - z^*\|}{\|z_k - z^*\|} < \infty$$

RF3

Proof.

$$\begin{aligned} z_{k+1} - z^* &= z_n - z^* - M_k^{-1} R(z_n) \\ &= z_n - z^* - M_k^{-1} [R(z_n) - R(z^*)] \\ &= M_k^{-1} \left[M_k (z_n - z^*) \right] - M_k^{-1} \int_0^1 (z^* + t(z_n - z^*)) (z_n - z^*) dt \\ &= M_k^{-1} \left[M_k (z_n - z^*) \right] (z_n - z^*) + \\ &\quad - M_k^{-1} \int_0^1 \left[\int (z^* + t(z_n - z^*)) - \int (z_n) \right] (z_n - z^*) dt \end{aligned}$$

Take the norm of both side, to get

$$\begin{aligned} \|z_{k+1} - z^*\| &\leq K_n \|z_n - z^*\| + \int_0^1 \|z^* + t(z_n - z^*) - z_n\| dt \|z_n - z^*\| \\ &= \left[K_n + \frac{1}{2} \int_0^1 (1-t) dt \|z_n - z^*\| \right] \|z_n - z^*\| \\ &= \left[K_n + \frac{1}{2} \|z_n - z^*\| \right] \|z_n - z^*\| \end{aligned}$$

□

THE LOCAL CONTRACTION THEOREM

Th. (Local contraction)

CONSIDER A NONLINEAR DIFFERENTIABLE FUNCTION

$R: \mathbb{R}^n \rightarrow \mathbb{R}^n$ AND A SOLUTION $\bar{z}^* \in \mathbb{R}^n$ (THAT IS, SUCH THAT $R(\bar{z}^*) = 0$)

CONSIDER A NEWTON'S TYPE OF ITERATION $\bar{z}_{k+1} = \bar{z}_k - M_k^{-1} R(\bar{z}_k)$
WITH INITIAL ITERATE \bar{z}_0

THE SEQUENCE $(\bar{z}_k)_{k=0}^\infty$ CONVERGES TO \bar{z}^* WITH CONTRACTION RATE

$$\|\bar{z}_{k+1} - \bar{z}^*\| \leq \underbrace{\left(K_k + \frac{w}{2} \|\bar{z}_k - \bar{z}^*\| \right)}_{\text{APPROXIMATE JACOBIAN}} \cdot \|\bar{z}_k - \bar{z}^*\|$$

OR ACTUAL JACOBIAN

IT GETS CLOSER
WITH $k \rightarrow \infty$

IF THERE EXIST $w < \infty$ AND $K < 1$ SUCH THAT THE FOLLOWING HOLDS

$$\begin{cases} \|M_k^{-1}[J(\bar{z}_k) - J(\bar{z})]\| \leq w \|\bar{z}_k - \bar{z}\| & (\text{Lipschitz condition}) \\ \|M_k^{-1}[J(\bar{z}_k) - M_k]\| \leq K_k < K & (\text{Compatibility condition}) \end{cases}$$

FOR ALL \bar{z}_k AND \bar{z} IN \mathbb{R}^n

AND, IF $\|\bar{z}_0 - \bar{z}^*\|$ IS SMALL ENOUGH, THAT IS SUCH THAT

$$\text{then } \|\bar{z}_0 - \bar{z}^*\| < \frac{2(1-K)}{w}$$

* DISTANCE BETWEEN JACOBIANS
** DISTANCE WITH APPROXIMATE

ITERATIONS START CLOSE ENOUGH TO THE SOLUTION

FOR $k=0$, THE METHOD IS THE EXACT NEWTON'S METHOD

————— // ————— //

IT COMPARES DISTANCES TO THE OPTIMAL SOLUTION
BETWEEN TWO SUCCESSIVE ITERATIONS
 \bar{z}_{k+1} AND \bar{z}_k

THE IDEA BEHIND THE CONTRACTION RATE

- Consider the two-dimensional case :

THE FIRST CONTRACTION FACTOR

$$\left(K_0 + \frac{w}{2} \| z_k - z^* \| \right) \text{ IS STRONGER THAN}$$

$$\left(K + \frac{w}{2} \| z_k - z^* \| \right) < 1$$

$$\text{BECAUSE } \| z_0 - z^* \| < \frac{2(1-K)}{w}$$

THIS IMPLIES THAT $\| z_1 - z^* \| \leq \delta \| z_0 - z^* \|$ WITH

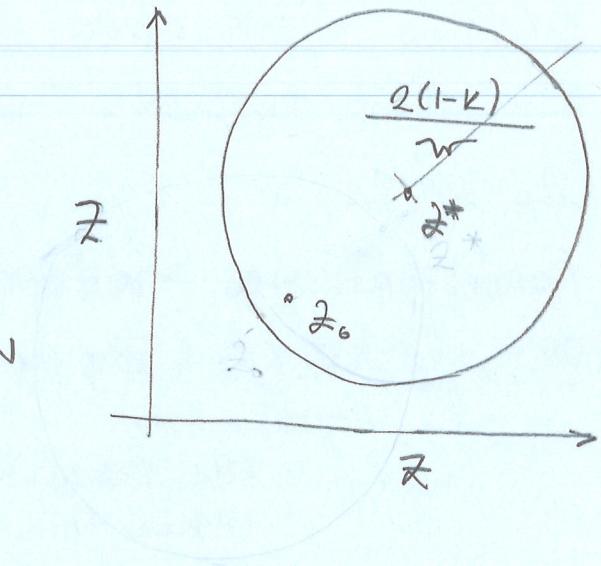
$$\delta = \left(K + \frac{w}{2} \| z_k - z^* \| \right)$$

RECURSIVELY SO FOR ALL CONTRACTION FACTORS WE HAVE THE BOUND δ SUCH THAT

$$\| z_k - z^* \| \leq \underbrace{\delta_K}_{\delta} \| z_0 - z^* \|$$

This implies linear convergence with contraction rate δ

LOCAL CONTRACTION RATES WILL BE USUALLY FASTER THAN THIS



- $R(z) = 0$ IS THE GIVEN PROBLEM
- A METHOD PRODUCES THE FOLLOWING SEQUENCE OF ITERATES
 z_0, z_1, z_2, \dots
- TRANSFORM THE PROBLEM (COORDINATE CHANGE)
 $\tilde{R}(y) = \underbrace{AR(b + By)}_{{\text{THE TRANSFORMED RESIDUAL}}_{\text{TAKES THIS FORM}}} = 0$
- APPLY THE METHOD TO $\tilde{R}(y) = 0$ FROM $y_0 = B^{-1}(z_0 - b)$
 (THE SAME INITIAL CONDITION IN THE NEW COORDINATE SYSTEM)
 $\rightarrow y_0, y_1, y_2, \dots$ SUCH THAT $y_k = B^{-1}(z_k - b) \quad \forall k$

From $z^* = b + By^*$

$\rightsquigarrow By^* = z^* - b$

$\rightsquigarrow B^{-1}By^* = B^{-1}(z^* - b)$

↓
AFFINE INVARIANCE
(CONDITIONS)

$$z_{k+1} = z_k - J(z_k)^{-1} R(z_k)$$

Does this update relate to $y_{k+1} = y_k - \tilde{J}(y_k)^{-1} \tilde{R}(y_k)$

We obtain,

$$y_{k+1} = y_k - B^{-1} J(b + By) \underbrace{\underbrace{A^{-1} A}_{I} R(b + By)}_{J(y_{k+1})^{-1} \tilde{R}(z_k)}$$

$$\tilde{J}(y_k) = A J(b + By) B$$

AFFINE INVARIANCE (of an algorithm)

Consider an iterative algorithm for rootfinding tasks $R(z) = 0$

→ THE METHOD IS CALLED AFFINE INVARIANT, IF AFFINE BASIC TRANSFORMATIONS OF THE EQUATIONS OF THE VARIABLES WILL NOT CHANGE THE RESULTING ITERATIONS

Consider 2 invertible matrices A and $B \in \mathbb{R}^{n \times n}$ and a vector $b \in \mathbb{R}^n$

↪ → CONSIDER THE FOLLOWING ROOT FINDING PROBLEM

$$\tilde{R}(y) = A \underbrace{R(b + By)}_{y^*} = 0 \quad \rightarrow z = By + b$$

→ THE SOLUTION TO $\tilde{R}(z) = 0$ IS z^*

If we have the solution z^* we can construct it from y^* such that $\tilde{R}(y^*) = 0$, by inverting the relation $z^* = b + By^*$

$$\rightarrow \text{THAT IS } y^* = B^{-1}(z^* - b)$$

↪ → CONSIDER AN ITERATIVE ALGORITHM THAT STARTING FROM THE INITIAL SOLUTION, x_0 , GENERATES THE ITERATES z_0, z_1, \dots THAT CONVERGE TO THE SOLUTION OF $R(z) = 0$

The method is called AFFINE INVARIANT IF, WHEN APPLIED TO $\tilde{R}(y) = 0$ from the initial solution $y_0 = B^{-1}(z_0 - b)$, the resulting iterates y_0, y_1, \dots all satisfy $y_k = B^{-1}(z_k - b)$

EXACT NEWTON's METHOD IS AFFINE INVARIANT, AS MANY OTHERS

AFFINE INVARIANCE IS AN IMPORTANT PROPERTY OF ITERATIVE METHODS FOR ROOT FINDING (no NUMERICAL OPTIMISATION)

SUPPOSE THAT YOUR TASK IS TO FIND THE EQUILIBRIUM TEMPERATURE IN A CHEMICAL REACTION SYSTEM

→ The problem can be formulated by considering temperatures expressed in different units (K, C, and F)

→ K, C, F formulations will lead to different numerical results corresponding to the same physical temperature

→ The different solutions can be obtained by affine transformations from one another

