

NEWTON-Algorithm (Basics)

1

hier: $\sqrt{2}$

2 1.5 1.41421 1.414215

$2, \frac{1}{2}\left[2 + \frac{2}{2}\right], \frac{1}{2}\left[1.5 + \frac{2}{1.5}\right], \frac{1}{2}\left[1.416 + \frac{2}{1.416}\right]$

$x^2 = 2 \quad x = \frac{2}{x} \quad \left. \begin{array}{l} x = \frac{2}{x} \\ x = x \end{array} \right\} \cdot \frac{1}{2} \quad x = \frac{1}{2}\left(x + \frac{2}{x}\right)$

$\sqrt{9} = 3 \quad -6 \quad x = \frac{9}{x} \quad \left. \begin{array}{l} x = \frac{9}{x} \\ x = x \end{array} \right\} \quad x = \frac{1}{2}\left(x + \frac{9}{x}\right)$

Kepler 1) nicht konvergiert

2) Man kann nicht sicher sein $\sqrt[3]{9} = ?$

2) -twz: KEPLER $x^3 = 9 \quad \left. \begin{array}{l} x^3 = 9 \\ x = x \end{array} \right\} \cdot \frac{1}{3} \quad x = \frac{2}{3}x + \frac{1}{3}\frac{9}{x^2}$

Seit's $\sqrt[3]{9} = 1.44225$ $x_0 := 9$

$x_{n+1} := \left(1 - \frac{1}{N}\right)x_n + \frac{1}{N}\frac{9}{x_n^{N-1}}$

1) NEWTON Algorithm $f(x) = 0$ [PE $\tilde{x} = 2 = 0$]

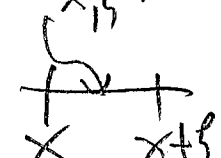


Erweitern - h/1000

TAYLOR ΕΡΗΥΝΑ + LAGRANGE ΜΑΡΑΔΕΚΤΑ

(3)

$$f(x+h) = f(x) + f'(x)h + \frac{1}{2} f''(x)h^2 + \dots + \frac{1}{N!} f^{(N)}(x)h^N + \frac{1}{(N+1)!} f^{(N+1)}(\xi_{x,h})h^{N+1}$$



ΗΑ f N-στήν
 ΕΥΤ ΔΙΦ

N=2

$$f(x+h) = f(x) + f'(x)h + \frac{1}{2} f''(\xi_{x,h})h^2$$

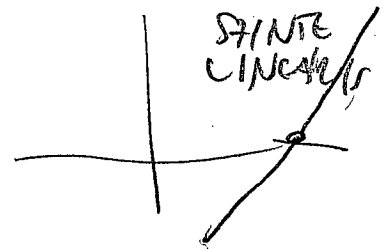
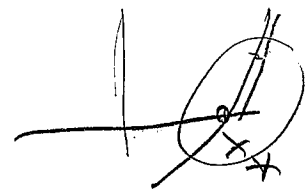
NEWTON ΜΕΥΡΙΑΤΗΛΩΣΤΑ

$$x_n \approx x_* \text{ ————— } f(x_*) = 0$$

$$f(x_n+h) \approx f(x_n) + f'(x_n)h$$

$$0 = f(x_*) \approx f(x_n) + f'(x_n)(x_* - x_n)$$

$$x_* \approx x_n - \frac{f(x_n)}{f'(x_n)}$$



RECUR

$$0 = f(x_*) = f(x_n) + f'(x_n)(x_* - x_n) + \frac{1}{2} f''(\xi_{x_n, x_*})(x_* - x_n)^2$$

$$0 = f(x_{n+1}) = f(x_n) + f'(x_n)(x_{n+1} - x_n)$$

$$(2) - (1) \Rightarrow$$

$$0 = f'(x_n)(x_{n+1} - x_*) - \frac{1}{2} f''(\xi_{n,x_*})(x_n - x_*)^2$$

$$|x_{n+1} - x_*| = \frac{1}{2} \frac{|f''(\xi_{n,x_*})|}{|f'(x_n)|} |x_n - x_*|^2$$

$$\leq M \quad \text{w.} \quad x_n \in (x_* - \varepsilon, x_* + \varepsilon)$$

$$\frac{1}{2} \frac{|f''(\xi)|}{|f'(\xi)|}$$

$$\exists \varepsilon > 0$$

$$\rho_n := |x_n - x_*|$$

x_* ist die Nullstelle

$$\boxed{\rho_{n+1} \leq M \rho_n^2}$$

DE M eine Konstante ist

$$\rho_1 \leq M \rho_0^2$$

— 2

$$\rho_2 \leq M \rho_1^2 \leq M(M \rho_0^2) = M^3 \rho_0^4$$

— 4

$$\rho_3 \leq M \rho_2^2 \leq M(M^3 \rho_0^4) = M^7 \rho_0^8$$

— 8

⋮

— 16

$$\rho_{n+1} \leq M^{2^n - 1} \rho_0^{2^n} = \frac{1}{M} [M \rho_0]^2$$

HA MDR $M \rho_0 < 1$ (AZA x_0 NACH KLEINER VON x_*)

AKKOR $x_n \rightarrow x_*$ mit $|x_n - x_*| \leq \text{const.} \cdot (1 - \delta)^{2^n}$

ÜBUNG 10: NEWTON-RAPHSON ITERATION

(5)

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

$$x_i = \begin{pmatrix} x_{i1} \\ \vdots \\ x_{id} \end{pmatrix} \mapsto f_i \quad \text{Koordinaten } f = \begin{pmatrix} f_1 \\ \vdots \\ f_k \end{pmatrix}$$

$$\nabla f_i(x) = \begin{bmatrix} \frac{\partial f_i}{\partial x_1} & \frac{\partial f_i}{\partial x_2} & \dots & \frac{\partial f_i}{\partial x_k} \end{bmatrix} \quad \text{GRADIENTEN (Spaltenvektor)}$$

$$\nabla^2 f_i(x) = \begin{bmatrix} \frac{\partial^2 f_i}{\partial x_h \partial x_l} \end{bmatrix}_{h,l=1}^k \quad \text{HESSE-MATRIXEN}$$

$$f_i(x+h) = f_i(x) + \langle \nabla f_i(x), h \rangle + \frac{1}{2} h^T \nabla^2 f_i(x) h$$

Teilhilf für x_h ADOTT (u. erweiterte x_{opt} -NAHE)
 $f(x_h) = 0$

$$f_i \approx [f_i]_h(x) := f_i(x_h) + \langle \nabla f_i(x_h), x - x_h \rangle$$

$$[f_i]_h(x) = 0 \quad (i=1, 2, \dots, k) \quad \text{GLOBALE MINIMIERUNG}$$

$$p \in \mathbb{R}^{k-2} \quad x = \begin{bmatrix} x \\ y \end{bmatrix} \quad f = \begin{bmatrix} f \\ g \end{bmatrix}$$

$$f(x_h) + \frac{\partial f}{\partial x}(x_h)(x - x_h) + \frac{\partial f}{\partial y}(y_h)(y - y_h) = 0$$

$$g(x_h) + \frac{\partial g}{\partial x}(x_h)(x - x_h) + \frac{\partial g}{\partial y}(y_h)(y - y_h) = 0$$

$$\underline{f}(\underline{x}) + \underbrace{\begin{bmatrix} \nabla f_1 \\ \vdots \\ \nabla f_k \end{bmatrix}}_{2 \times 2 \text{-s matrix}} (\underline{x}_{n+1} - \underline{x}_n) = 0$$

2x2-s matrix is JACOBI MATRIX

Algorithm is: (Take $k \geq 1$ DIM-BASED)

$$\underline{f}(\underline{x}_n) + \underbrace{\begin{bmatrix} \nabla f_1(\underline{x}_n) \\ \vdots \\ \nabla f_k(\underline{x}_n) \end{bmatrix}}_{k \times k \text{-s matrix}} (\underline{x}_{n+1} - \underline{x}_n) = 0$$

MEGOLDAND $\underline{x}_{n+1}$ -RE

$$\underline{x}_{n+1} := \underline{x}_n - \underbrace{\begin{bmatrix} \nabla f_1(\underline{x}_n) \\ \vdots \\ \nabla f_k(\underline{x}_n) \end{bmatrix}^{-1}}_{\underline{f}'(\underline{x}_n)} \underline{f}(\underline{x}_n)$$

Besides

$$\underline{x}_{n+1} - \underline{x}_* = \frac{1}{2} \underline{f}'(\underline{x}_n)^{-1} \begin{bmatrix} (\underline{x}_n - \underline{x}_*)^T \nabla^2 f_1(\underline{x}_n) (\underline{x}_n - \underline{x}_*) \\ \vdots \\ (\underline{x}_n - \underline{x}_*)^T \nabla^2 f_k(\underline{x}_n) (\underline{x}_n - \underline{x}_*) \end{bmatrix}$$

$$\|\underline{x}_{n+1} - \underline{x}_*\| \leq M \|\underline{x}_n - \underline{x}_*\|^2$$