

Summary of Newton's method to solve  $f(x) = 0$ ,  $f: [a, b] \rightarrow \mathbb{R}$

Choose  $x_0 \in [a, b]$  (initialization)

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} = x_k - f'(x_k)^{-1} f(x_k) \quad k=0, 1, 2, \dots$$

Theorem: (Convergence of Newton's method)

$f$  cont,  $f''$  continuous on  $I_\delta = [\xi - \delta, \xi + \delta]$ ,  $\delta > 0$

$f(\xi) = 0$ ,  $f'(\xi) \neq 0$ ,  $\exists A > 0$ :

$$\left| \frac{f''(x)}{f'(y)} \right| \leq A \quad \text{for all } x, y \in I_\delta$$

If  $|\xi - x_0| \leq h := \min(\delta, 1/A)$ , then the Newton sequence

$x_k \rightarrow \xi$  quadratically.

$$\frac{|x_{k+1} - \xi|}{|x_k - \xi|^2} \rightarrow \eta < \infty$$

Remark: implies that  $f'(\xi) \neq 0$ .

Newton's method requires  $f'(x_k)$ , in every iteration.

Example: Newton to solve  $f(x) = x^3 = 0$

Obvious solution  $x=0$

§ 1.5 Secant method

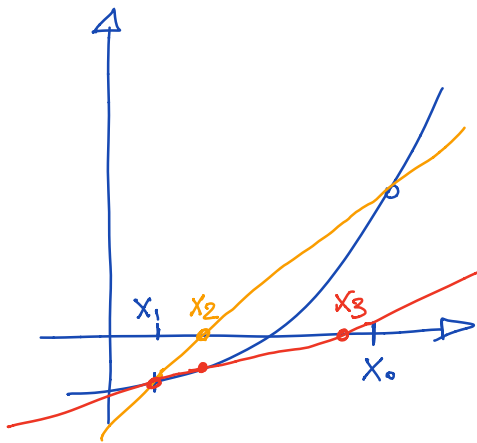
does not require  $f'$ , as it approximates derivatives by difference quotients:

$$f'(x_k) \approx \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}} \quad \text{if } x_k \neq x_{k-1}$$

Secant method:

$$x_0, x_1 \text{ initialization} \quad \approx f'(x_k)^{-1}$$

$$x_{k+1} = x_k - \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} f(x_k), \quad k=1, 2, 3, \dots$$



Theorem:  $f$  continuously differentiable on  $[\xi-h, \xi+h]$ ,  $h > 0$   
 $f(\xi) = 0$ ,  $f'(\xi) \neq 0$

Then the secant method converges at least linearly to  $\xi$  if  $x_0, x_1$  are sufficiently close to  $\xi$ .

Remark: One can show that the secant method converges faster:

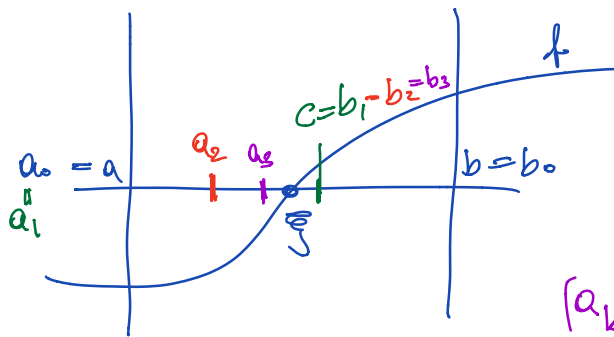
$$\lim_{k \rightarrow \infty} \frac{|x_{k+1} - \xi|}{|x_k - \xi|^2} = \mu < \infty$$

$$\text{for } \rho = \frac{1}{2} (1 + \sqrt{5}) \approx \underline{\underline{1.6}}$$

The method is slower than Newton's method, but 'cheaper' since I do not require  $f'(x_k)$ .

## § 1.6. Bisection method

$f: [a, b] \rightarrow \mathbb{R}$ ,  $\xi \in [a, b]$   $f(\xi) = 0$ ,  $f$  continuous.



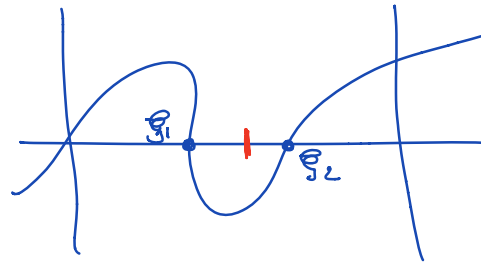
$(a_0, b_0)$   
check if there is a  
sign change between  
 $a_0, c$  or  $c, b_0$

$$(a_{k+1}, b_{k+1}) = \begin{cases} (a_k, c_k) & \text{if } f(a_k)f(b_k) > 0 \\ (c_k, b_k) & \text{else} \end{cases}$$

- Simple, robust
- slower than Newton's method, at least asymptotically.
- issues if there are several roots of  $f$ .

$$c_k = \frac{a_k + b_k}{2}$$

depending on how  
we choose the  
new  $(a_k, b_k)$ , we  
converge to either  
 $\xi_1$  or  $\xi_2$



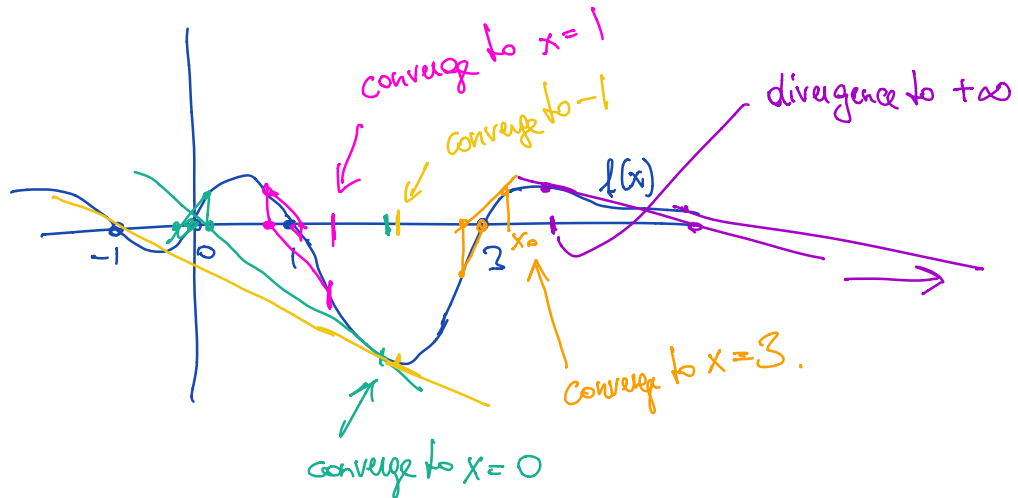
## § 1.7 Global behavior of Newton's method

Convergence if starting point  $x_0$  is close to  $\xi_i$ ; general behavior can be very complicated

Example:  $f(x) = x(x^2 - 1)(x - 3) \exp\left(-\frac{1}{2}(x-1)^2\right)$

roots:

$$\begin{aligned} x &= 0 \\ x &= \pm 1 \\ x &= 3 \end{aligned}$$



Newton's method for systems of equations

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}^2 \quad f(x,y) = \begin{pmatrix} f_1(x,y) \\ f_2(x,y) \end{pmatrix}$$

Look for  $(\xi_1, \xi_2) \in \mathbb{R}^2 : f(\xi_1, \xi_2) = 0$

Newton's method: Choose  $\begin{pmatrix} x_0 \\ y_0 \end{pmatrix} \in \mathbb{R}^2$  initialization

$$\begin{pmatrix} x_{k+1} \\ y_{k+1} \end{pmatrix} = \begin{pmatrix} x_k \\ y_k \end{pmatrix} - \underbrace{Df(x_k, y_k)^{-1}}_{\in \mathbb{R}^{2 \times 2}} \underbrace{f(x_k, y_k)}_{\in \mathbb{R}^2} \quad k=0, 1, 2, \dots$$

$$Df(x,y) = \begin{pmatrix} \frac{\partial f_1}{\partial x} & \frac{\partial f_1}{\partial y} \\ \frac{\partial f_2}{\partial x} & \frac{\partial f_2}{\partial y} \end{pmatrix} \in \mathbb{R}^{2 \times 2}$$

Example:  $f(x,y) = \begin{pmatrix} x^2 + y^2 - 2 \\ 2x^2 - y^2 - 1 \end{pmatrix} = \begin{pmatrix} f_1(x,y) \\ f_2(x,y) \end{pmatrix}$

solution is  $(\beta_1, \beta_2) = (1, 1)$ , also:  $(-1, 1)$ ,  $(1, -1)$ ,  
 $(-1, -1)$

$$Df(x, y) = \begin{pmatrix} 2x & 2y \\ 4x & -2y \end{pmatrix}$$

Newton's Start  $\begin{pmatrix} x_0 \\ y_0 \end{pmatrix}$

$$\begin{pmatrix} x_{k+1} \\ y_{k+1} \end{pmatrix} = \begin{pmatrix} x_k \\ y_k \end{pmatrix} - \begin{pmatrix} 2x_k & 2y_k \\ 4x_k & -2y_k \end{pmatrix}^{-1} \begin{pmatrix} x_k^2 + y_k^2 - 2 \\ 2x_k^2 - y_k^2 - 1 \end{pmatrix}$$

- Remarks:
- requires the Jacobian matrix ("expensive")
  - requires matrix inverse  
(or to solve a linear system)
  - similar convergence results hold, i.e.,  
"quadratic convergence close to solution"