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Chapter 3 Solution of Nonlinear Equations

§3.0 Introduction $f: \mathbb{R} \rightarrow \mathbb{R}$

Find $x \in \mathbb{R}$ s.t. $f(x) = 0$.

methods: bisection, Newton, secant

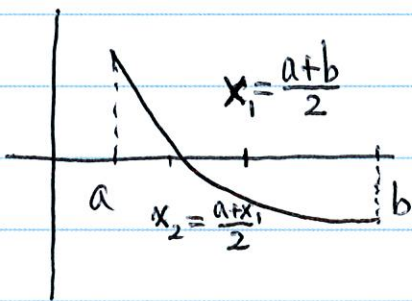
theory: fixed-pt, continuation

zeros of polynomials

§3.1 Bisection (Interval Halving) Method

Assumption $f \in C[a, b]$ and $f(a)f(b) < 0$

$\Rightarrow \exists r \in (a, b)$ s.t. $f(r) = 0$



stopping criterion $|f(x_n)| < 10^{-5}$

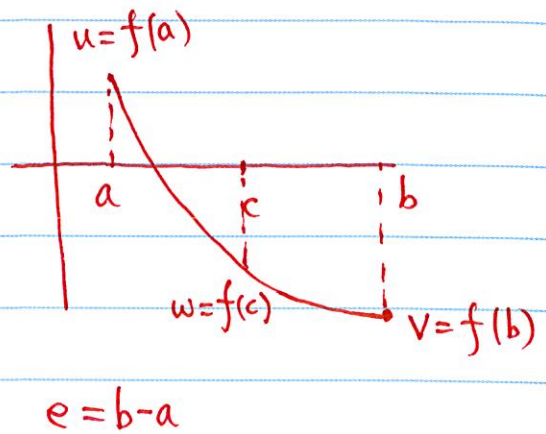
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Example Find x s.t. $e^x = \sin x$ closest to 0.

$f(x) = e^x - \sin x$ $[a, b] = [-4, -3]$ (P76)

Bisection Algorithm

input $a, b, M, \delta, \epsilon$
 $u \leftarrow f(a)$
 $v \leftarrow f(b)$
 $e \leftarrow b - a$



output a, b, u, v

if $\text{sign}(u) = \text{sign}(v)$ then stop
 for $k = 1$ to M do

$e \leftarrow \frac{e}{2}$

$c \leftarrow a + e$ $c = a + (b - a) / 2$

$w \leftarrow f(c)$

output k, c, w, e

if $|e| < \delta$ or $|w| < \epsilon$ then stop

if $\text{sign}(w) \neq \text{sign}(u)$ then

$b \leftarrow c$

$v \leftarrow w$

else

$a \leftarrow c$

$u \leftarrow w$

end if

end do

$\frac{a+b}{2}$ could moves outside of $[a, b]$ on a limited precision machine

$uv < 0$
unnecessary multiplication overflow or underflow

3 stopping criterion:
 M, δ, ϵ

Error Analysis

$$[a_0, b_0] = [a, b] \implies \lim_{n \rightarrow \infty} a_n \text{ exists}$$

$$\implies a_0 \leq a_1 \leq a_2 \dots \leq b_0 \implies b_{n+1} - a_{n+1} = \frac{1}{2}(b_n - a_n)$$

$$b_0 \geq b_1 \geq b_2 \dots \geq a_0 \implies \lim_{n \rightarrow \infty} b_n \text{ exists}$$

$$b_n - a_n = \frac{1}{2}(b_{n-1} - a_{n-1}) = \frac{1}{2^n}(b_0 - a_0)$$

$$\implies \lim_{n \rightarrow \infty} b_n = \lim_{n \rightarrow \infty} a_n = r$$

$$0 \geq f(a_n)f(b_n) \implies 0 \geq [f(r)]^2 \implies f(r) = 0$$

Let $c_n = \frac{1}{2}(a_n + b_n)$

$$|r - c_n| \leq \frac{1}{2}(b_n - a_n) = \frac{1}{2^{n+1}}(b_0 - a_0)$$

Thrm on Bisection Method

Let $[a_n, b_n]$ be the intervals in the bisection method

$$\implies (1) \lim_{n \rightarrow \infty} a_n = \lim_{n \rightarrow \infty} b_n = r \text{ and } f(r) = 0$$

$$(2) \left| r - \frac{1}{2}(a_n + b_n) \right| \leq \frac{1}{2^{n+1}}(b_0 - a_0)$$

example $[a, b] = [50, 63]$, find n st. $\left| \frac{r - c_n}{r} \right| \leq 10^{-12}$

$$\leq \frac{|r - c_n|}{50} \leq \frac{1}{50} \cdot \frac{1}{2^{n+1}}(63 - 50)$$

$$\implies n \geq 37$$

§3.2 Newton's Method (Newton-Raphson Iteration)

Assumption $f''(x)$ exists and is cont. and $f(r) = 0$

Let x be an approx to $r \Rightarrow r \approx x + h$

$$0 = f(r) = f(x+h) = f(x) + hf'(x) + O(h^2)$$

$$\approx f(x) + hf'(x) \quad \text{if } h = r - x \text{ is small}$$

$$\Rightarrow h = -\frac{f(x)}{f'(x)}$$

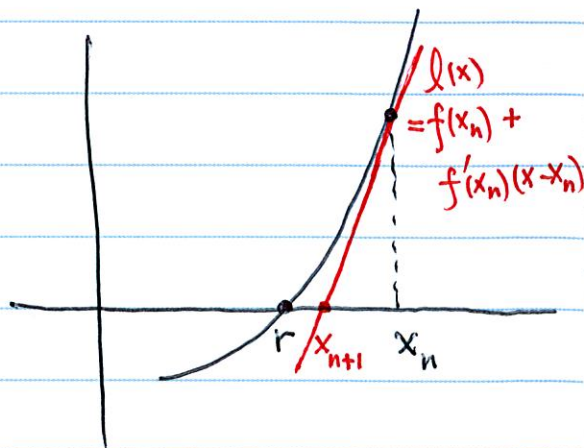
$$\Rightarrow x+h = x - \frac{f(x)}{f'(x)} \quad \text{is a better approx than } x$$

or $0 = f(r) = f(x) + f'(x)(r-x) + O((r-x)^2)$

$$\approx f(x) + f'(x)(r-x)$$

$$\Rightarrow r \approx x - f(x)/f'(x)$$

$$\boxed{x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}}$$



Newton's Algorithm

input x, M

$y \leftarrow f(x)$

output $0, x, y$

for $k=1$ to M do

$x \leftarrow x - y/f'(x)$

$y \leftarrow f(x)$

output k, x, y

end do

~~or~~

Or input x_0, M, δ, ϵ

$v \leftarrow f(x_0)$

output $0, x_0, v$

if $|v| < \epsilon$ then stop

for $k=1$ to M do

$x_1 \leftarrow x_0 - v/f'(x_0)$

$v \leftarrow f(x_1)$

output k, x_1, v

if $|x_1 - x_0| < \delta$ or $|v| < \epsilon$ then stop

$x_0 \leftarrow x_1$

end do

example $f(x) = e^x - 1.5 - \tan^{-1}x$, find the negative zero.

Error Analysis

Assumptions f'' is cont., $f(r)=0$, and $f'(r) \neq 0$

error $e_n = x_n - r$

$$e_{n+1} = x_{n+1} - r = \frac{e_n f'(x_n) - f(x_n)}{f'(x_n)}$$

$$0 = f(r) = f(x_n - e_n) = \boxed{f(x_n) - e_n f'(x_n)} + \frac{1}{2} e_n^2 f''(\xi_n)$$

$$\Rightarrow e_{n+1} = \frac{1}{2} \frac{f''(\xi_n)}{f'(x_n)} e_n^2 \approx \frac{1}{2} \frac{f''(r)}{f'(r)} e_n^2$$

Let $c(\delta) = \frac{1}{2} \frac{\max_{|x-r| \leq \delta} |f''(x)|}{\min_{|x-r| \leq \delta} |f'(x)|}$ for $\delta > 0$

$$\Rightarrow |e_1| = \frac{1}{2} \frac{|f''(\xi_0)|}{|f'(x_0)|} e_0^2 \leq c(\delta) e_0^2$$

choose small enough δ s.t.
 $\min_{|x-r| \leq \delta} |f'(x)| > 0$
and $\delta c(\delta) < 1$

$$\leq c(\delta) \delta |e_0|$$

if $|e_0| \leq \delta$

$$= \rho |e_0| < |e_0| \leq \delta$$

$$\rho = c(\delta) \delta$$

$$|e_1| \leq \delta$$

$$\Rightarrow |e_n| \leq \rho^n |e_0|$$

Thm 1 Assume that f'' is cont. and let r be a simple zero of f , $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$

$\Rightarrow \exists O(r, \delta)$, if $x_0 \in O(r, \delta)$, then $\lim_{n \rightarrow \infty} x_n = r$.

moreover, \exists const $c > 0$ s.t.

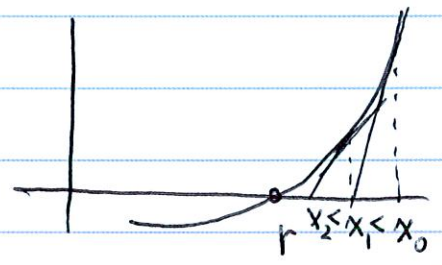
$$|x_{n+1} - r| \leq c (x_n - r)^2$$

Thm 2 (convex function)

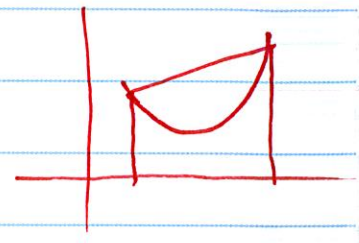
Assume that $f \in C^2(\mathbb{R})$, is increasing, is convex, and has a zero,
 $f'' > 0$ or $f(tx_1 + (1-t)x_2) \leq tf(x_1) + (1-t)f(x_2)$

- \Rightarrow (1) the zero is unique
- (2) the Newton iteration converges globally.

Proof



increasing $f'(x) > 0$
convex $f''(x) > 0$



$$x_{n+1} - r = \frac{1}{2} \frac{f''(\xi_n)}{f'(x_n)} e_n^2 > 0$$

$$\Rightarrow x_n > r \quad \forall n \geq 1 \xrightarrow{f' > 0} f(x_n) > f(r) = 0$$

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} < x_n \quad \text{and} \quad x_n > r \Rightarrow \lim_{n \rightarrow \infty} x_n = x^*$$

$$e_{n+1} = e_n - \frac{f(x_n)}{f'(x_n)} < e_n \quad \text{and} \quad e_n > 0 \Rightarrow \lim_{n \rightarrow \infty} e_n = e^*$$

$$\Rightarrow e^* = e^* - \frac{f(x^*)}{f'(x^*)} \Rightarrow \boxed{f(x^*) = 0}$$

$$\text{and } x^* = r$$

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Example Newton's method for $f(x) \equiv x^2 - R = 0$ for $R > 0$

$$x_{n+1} = \frac{1}{2} \left(x_n + \frac{R}{x_n} \right)$$

Implicit Functions (an application of Newton's method)

$$G(x, y(x)) = 0$$

for a prescribed x_i , } given y_0

find ~~y~~ y s.t. $f(y) = G(x_i, y) = 0$ $y_{k+1} = y_k - G(x_i, y_k) / \frac{\partial G}{\partial y}(x_i, y_k)$

$$y_{k+1} = y_k - \frac{f(y_k)}{f'(y_k)}$$

compute an approx to y_1 in which $G(x_i, y_1) = 0$

\Rightarrow generate a table for $y(x)$. (see p87)

Systems of Nonlinear Equations

$$\begin{cases} f_1(x_1, x_2) = 0 \\ f_2(x_1, x_2) = 0 \end{cases} \quad \text{or} \quad \begin{pmatrix} f_1(x_1, x_2) \\ f_2(x_1, x_2) \end{pmatrix} = \vec{0}$$

Let (x_1, x_2) be an approximation, and let (h_1, h_2) be the correction to be computed, then $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} h_1 \\ h_2 \end{pmatrix}$ new approximation

$$0 = f_1(x_1 + h_1, x_2 + h_2) \approx f_1(x_1, x_2) + h_1 \frac{\partial f_1}{\partial x_1}(x_1, x_2) + h_2 \frac{\partial f_1}{\partial x_2}(x_1, x_2)$$

$$0 = f_2(x_1 + h_1, x_2 + h_2) \approx f_2(x_1, x_2) + h_1 \frac{\partial f_2}{\partial x_1}(x_1, x_2) + h_2 \frac{\partial f_2}{\partial x_2}(x_1, x_2)$$

$$\Leftrightarrow \vec{0} = \vec{f}(\vec{x} + \vec{h}) \approx \vec{f}(\vec{x}) + \begin{bmatrix} \partial_1 f_1(\vec{x}) & \partial_2 f_1(\vec{x}) \\ \partial_1 f_2(\vec{x}) & \partial_2 f_2(\vec{x}) \end{bmatrix} \vec{h}$$

$$\Rightarrow \vec{h} = -J^{-1} \vec{f}(\vec{x})$$

$\begin{matrix} \text{J} \\ \text{Jacobían matrix} \end{matrix}$

Newton's Iteration

$$\vec{x}^{(k+1)} = \vec{x}^{(k)} - J^{-1}(\vec{x}^{(k)}) \vec{f}(\vec{x}^{(k)}) \quad \text{for } k=0,1,\dots$$

example 4

$$\begin{cases} xy = z^2 + 1 \\ xyz + y^2 = x^2 + 2 \\ e^x + z = e^y + 3 \end{cases}$$

$$\vec{f}(x) = \begin{pmatrix} x_1 x_2 - x_3^2 - 1 \\ x_1 x_2 x_3 + x_2^2 - x_1^2 - 2 \\ e^{x_1} + x_3 - e^{x_2} - 3 \end{pmatrix} = \vec{0}$$

$$J = \begin{pmatrix} \partial_1 f_1 & \partial_2 f_1 & \partial_3 f_1 \\ \partial_1 f_2 & \partial_2 f_2 & \partial_3 f_2 \\ \partial_1 f_3 & \partial_2 f_3 & \partial_3 f_3 \end{pmatrix} = \begin{pmatrix} x_2 & x_1 & -2x_3 \\ x_2 x_3 - 2x_1 & x_1 x_3 + 2x_2 & x_1 x_2 \\ e^{x_1} & -e^{x_1} & 1 \end{pmatrix}$$

$$\vec{x}^{(0)} = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

3.3 Secant Method

to avoid computing derivative in Newton's method several methods have been studied.

(1) Steffensen's iteration

$$x_{n+1} = x_n - \frac{f(x_n)^2}{f(x_n + f(x_n)) - f(x_n)}$$

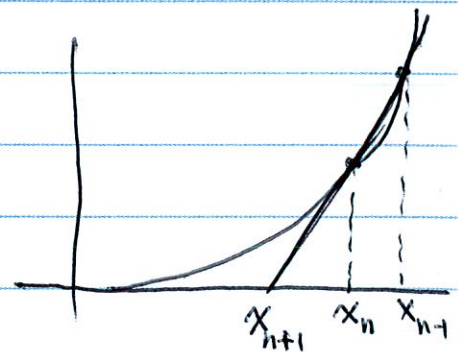
quadratically convergent

(*) • $f'(x) \approx g(x) = \frac{f(x + f(x)) - f(x)}{f(x)}$

(2) the secant method

$$x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}$$

• $f'(x_n) \approx \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$



example

$$f(x) = x^3 - \sinh x + 4x^2 + 6x + 9$$

there is a zero between $x_0 = 7$ and $x_1 = 8$

Secant Algorithm

input $a, b, M, \delta, \varepsilon$

$f_a \leftarrow f(a); f_b \leftarrow f(b)$

output $0, a, f_a$

output $1, b, f_b$

for $k=2$ to M do

if $|f_a| > |f_b|$ then

$a \leftrightarrow b; f_a \leftrightarrow f_b$

endif

$s \leftarrow (b-a)/(f_b-f_a)$

$b \leftarrow a$

$f_b \leftarrow f_a$

$a \leftarrow a - f_a * s$

$f_a \leftarrow f(a)$

output k, a, f_a

if $|f_a| < \varepsilon$ or $|b-a| < \delta$ then stop

enddo

$$\left[\begin{array}{cc} a & b \\ || & || \\ \{x_n, x_{n-1}\} & \\ |f(x_n)| \leq |f(x_{n-1})| & \\ \text{f}_a & \text{f}_b \end{array} \right]$$

$$x_n = x_n - f(x_n) * s$$

Error Analysis

$$x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} = \frac{f(x_n)x_{n-1} - f(x_{n-1})x_n}{f(x_n) - f(x_{n-1})}$$

Let $e_n = x_n - r$

$$\begin{aligned} \Rightarrow e_{n+1} &= x_{n+1} - r = \frac{f(x_n)e_{n-1} - f(x_{n-1})e_n}{f(x_n) - f(x_{n-1})} \\ &= \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \cdot \frac{\frac{f(x_n)}{e_n} - \frac{f(x_{n-1})}{e_{n-1}}}{x_n - x_{n-1}} \cdot e_n e_{n-1} \end{aligned}$$

Taylor's expansion

$$f(x_n) = f(e_n + r) = f(r) + e_n f'(r) + \frac{1}{2} e_n^2 f''(r) + O(e_n^3)$$

$$\Rightarrow \frac{f(x_n)}{e_n} = f'(r) + \frac{1}{2} e_n f''(r) + O(e_n^2)$$

$$\Rightarrow \frac{f(x_{n-1})}{e_{n-1}} = f'(r) + \frac{1}{2} e_{n-1} f''(r) + O(e_{n-1}^2)$$

$$\frac{f(x_n)}{e_n} - \frac{f(x_{n-1})}{e_{n-1}} = \frac{1}{2} (e_n - e_{n-1}) f''(r) + O(e_{n-1}^2)$$

$$\frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}} = \frac{(e_n - e_{n-1}) f'(r) + O(e_{n-1}^2)}{x_n - x_{n-1}} = \frac{1}{2} (x_n - x_{n-1}) f''(r) + O(e_{n-1}^2)$$

$$\Rightarrow e_{n+1} \approx \frac{1}{f'(r)} \cdot \frac{1}{2} f''(r) e_n e_{n-1} = C e_n e_{n-1}$$

Assume that $|e_{n+1}| \sim A |e_n|^\alpha$ i.e., $\frac{|e_{n+1}|}{A |e_n|^\alpha} \rightarrow 1$ as $n \rightarrow \infty$

$$\Rightarrow |e_n| \sim A |e_{n-1}|^\alpha \text{ or } |e_{n-1}| \sim (A^{-1} |e_n|)^{\frac{1}{\alpha}}$$

$$|e_{n+1}| \sim |c| |e_n| |e_{n-1}|$$

$$\underbrace{\quad}_A |e_n|^\alpha \quad \underbrace{\quad}_{|c| |e_n|} A^{-\frac{1}{\alpha}} |e_n|^{\frac{1}{\alpha}}$$

$$\Rightarrow |e_n|^{1-\alpha+\frac{1}{\alpha}} \sim A^{1+\frac{1}{\alpha}} |c|^{-1}$$

$$\Rightarrow 1-\alpha+\frac{1}{\alpha}=0 \Rightarrow \alpha = \frac{1+\sqrt{5}}{2} \approx 1.62 \quad \underline{\text{superlinear}}$$

$$A^{1+\frac{1}{\alpha}} |c|^{-1} \sim 1 \Rightarrow A = |c|^{\frac{\alpha}{1+\alpha}} = |c|^{0.62} \approx \left| \frac{f''(r)}{2f'(r)} \right|^{0.62}$$

Comparison with Newton's method

Newton 2 evaluation $f(x_n), f'(x_n) \Rightarrow$ 1 step Newton \approx 2 steps Secant
 Secant 1 evaluation $f(x_n)$

$$|e_{n+2}| \sim A |e_{n+1}|^\alpha \sim A^{1+\alpha} |e_n|^{\alpha^2} = A^{1+\alpha} |e_n|^{\frac{3+\sqrt{5}}{2}} \approx 2.62$$