# Numerical Analysis I Solutions of Equations in One Variable

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# Outline

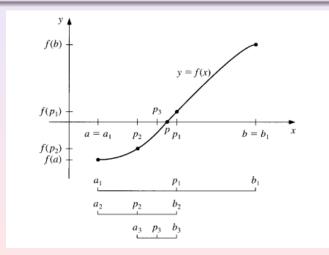
- Bisection Method
- 2 Fixed-Point Iteration
- Newton's method
- 4 Error analysis for iterative methods
- 5 Accelerating convergence
- 6 Zeros of polynomials and Müller's method



# **Bisection Method**

#### Idea

If  $f(x) \in C[a, b]$  and f(a)f(b) < 0, then  $\exists c \in (a, b)$  such that f(c) = 0.





#### Bisection method algorithm

Given f(x) defined on (a, b), the maximal number of iterations M, and stop criteria  $\delta$  and  $\varepsilon$ , this algorithm tries to locate one root of f(x).

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Compute fa = f(a), fb = f(b), and e = b - a
If sign(fa) = sign(fb), then stop End If
For k = 1, 2, \dots, M
  e = e/2, c = (a + b)/2, fc = f(c)
  If (|e| < \delta \text{ or } |fc| < \varepsilon), then stop End If
  If sign(fc) \neq sign(fa)
     b = c, fb = fc
   Else
     a = c. fa = fc
   End If
End For
```

Let  $\{c_n\}$  be the sequence of numbers produced. The algorithm should stop if one of the following conditions is satisfied.

- **1** the iteration number k > M,
- **2**  $|c_k c_{k-1}| < \delta$ , or
- $|f(c_k)| < \varepsilon.$

Let  $[a_0, b_0], [a_1, b_1], \cdots$  denote the successive intervals produced by the bisection algorithm. Then

$$a = a_0 \le a_1 \le a_2 \le \cdots \le b_0 = b$$
  
 $\Rightarrow \{a_n\} \text{ and } \{b_n\} \text{ are bounded}$   
 $\Rightarrow \lim_{n \to \infty} a_n \text{ and } \lim_{n \to \infty} b_n \text{ exist}$ 



Since

$$b_1 - a_1 = \frac{1}{2}(b_0 - a_0)$$

$$b_2 - a_2 = \frac{1}{2}(b_1 - a_1) = \frac{1}{4}(b_0 - a_0)$$

$$\vdots$$

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

hence

$$\lim_{n\to\infty}b_n-\lim_{n\to\infty}a_n=\lim_{n\to\infty}(b_n-a_n)=\lim_{n\to\infty}\frac{1}{2^n}(b_0-a_0)=0.$$

Therefore

$$\lim_{n\to\infty}a_n=\lim_{n\to\infty}b_n\equiv z.$$

Since f is a continuous function, we have that

$$\lim_{n\to\infty} f(a_n) = f(\lim_{n\to\infty} a_n) = f(z) \quad \text{and} \quad \lim_{n\to\infty} f(b_n) = f(\lim_{n\to\infty} b_n) = f(z).$$

On the other hand,

$$f(a_n)f(b_n) < 0$$

$$\Rightarrow \lim_{n \to \infty} f(a_n)f(b_n) = f^2(z) \le 0$$

$$\Rightarrow f(z) = 0$$

Therefore, the limit of the sequences  $\{a_n\}$  and  $\{b_n\}$  is a zero of f in [a,b]. Let  $c_n=\frac{1}{2}(a_n+b_n)$ . Then

$$|z - c_n| = \left| \lim_{n \to \infty} a_n - \frac{1}{2} (a_n + b_n) \right|$$

$$= \left| \frac{1}{2} \left[ \lim_{n \to \infty} a_n - b_n \right] + \frac{1}{2} \left[ \lim_{n \to \infty} a_n - a_n \right] \right|$$

$$\leq \max \left\{ \left| \lim_{n \to \infty} a_n - b_n \right|, \left| \lim_{n \to \infty} a_n - a_n \right| \right\}$$

$$\leq |b_n - a_n| = \frac{1}{2^n} |b_0 - a_0|.$$

This proves the following theorem.



#### **Theorem**

Let  $\{[a_n, b_n]\}$  denote the intervals produced by the bisection algorithm. Then  $\lim_{n\to\infty} a_n$  and  $\lim_{n\to\infty} b_n$  exist, are equal, and represent a zero of f(x). If

$$z = \lim_{n \to \infty} a_n = \lim_{n \to \infty} b_n$$
 and  $c_n = \frac{1}{2}(a_n + b_n),$ 

then

$$|z-c_n|\leq \frac{1}{2^n}(b_0-a_0).$$

#### Remark

 $\{c_n\}$  converges to z with the rate of  $O(2^{-n})$ .



### Example

How many steps should be taken to compute a root of  $f(x) = x^3 + 4x^2 - 10 = 0$  on [1,2] with relative error  $10^{-3}$ ?

solution: Seek an n such that

$$\frac{|z-c_n|}{|z|} \le 10^{-3} \Rightarrow |z-c_n| \le |z| \times 10^{-3}.$$

Since  $z \in [1, 2]$ , it is sufficient to show

$$|z-c_n|\leq 10^{-3}.$$

That is, we solve

$$2^{-n}(2-1) \le 10^{-3} \Rightarrow -n \log_{10} 2 \le -3$$

which gives  $n \ge 10$ .



### Fixed-Point Iteration

#### **Definition**

x is called a fixed point of a given function f if f(x) = x.

# Root-finding problems and fixed-point problems

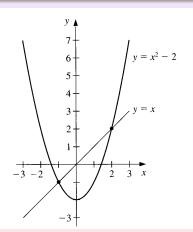
- Find  $x^*$  such that  $f(x^*) = 0$ . Let g(x) = x - f(x). Then  $g(x^*) = x^* - f(x^*) = x^*$ .  $\Rightarrow x^*$  is a fixed point for g(x).
- Find  $x^*$  such that  $g(x^*) = x^*$ . Define f(x) = x - g(x) so that  $f(x^*) = x^* - g(x^*) = x^* - x^* = 0$  $\Rightarrow x^*$  is a zero of f(x).



#### Example

The function  $g(x) = x^2 - 2$ , for  $-2 \le x \le 3$ , has fixed points at x = -1 and x = 2 since

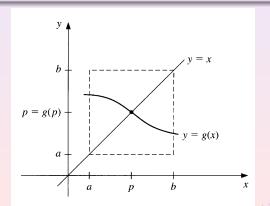
$$g(-1) = (-1)^2 - 2 = -1$$
 and  $g(2) = 2^2 - 2 = 2$ .





# Theorem (Existence and uniqueness)

- If  $g \in C[a, b]$  such that  $a \leq g(x) \leq b$  for all  $x \in [a, b]$ , then g has a fixed point in [a, b].
- 2 If, in addition, g'(x) exists in (a, b) and there exists a positive constant M < 1 such that  $|g'(x)| \leq M < 1$  for all  $x \in (a, b)$ . Then the fixed point is unique.







# **Proof**

#### Existence:

- If g(a) = a or g(b) = b, then a or b is a fixed point of g and we are done.
- Otherwise, it must be g(a) > a and g(b) < b. The function h(x) = g(x) x is continuous on [a, b], with

$$h(a) = g(a) - a > 0$$
 and  $h(b) = g(b) - b < 0$ .

By the Intermediate Value Theorem,  $\exists \ x^* \in [a,b]$  such that  $h(x^*) = 0$ . That is

$$g(x^*) - x^* = 0 \implies g(x^*) = x^*.$$

Hence g has a fixed point  $x^*$  in [a, b].



# Proof

#### Uniqueness:

Suppose that  $p \neq q$  are both fixed points of g in [a, b]. By the Mean-Value theorem, there exists  $\xi$  between p and q such that

$$g'(\xi) = \frac{g(p) - g(q)}{p - q} = \frac{p - q}{p - q} = 1.$$

However, this contradicts to the assumption that  $|g'(x)| \leq M < 1$  for all x in [a, b]. Therefore the fixed point of g is unique.





### Example

Show that the following function has a unique fixed point.

$$g(x) = (x^2 - 1)/3, x \in [-1, 1].$$

Solution: The Extreme Value Theorem implies that

$$\min_{x \in [-1,1]} g(x) = g(0) = -\frac{1}{3}, 
\max_{x \in [-1,1]} g(x) = g(\pm 1) = 0.$$

That is  $g(x) \in [-1, 1], \ \forall \ x \in [-1, 1].$ 

Moreover, g is continuous and

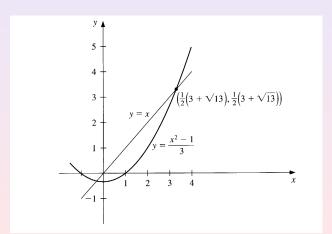
$$|g'(x)| = \left|\frac{2x}{3}\right| \le \frac{2}{3}, \ \forall \ x \in (-1,1).$$

By above theorem, g has a unique fixed point in [-1,1].



Let p be such unique fixed point of g. Then

$$p = g(p) = \frac{p^2 - 1}{3} \quad \Rightarrow \quad p^2 - 3p - 1 = 0$$
$$\Rightarrow \quad p = \frac{1}{2}(3 - \sqrt{13}).$$







### Fixed-point iteration or functional iteration

Given a continuous function g, choose an initial point  $x_0$  and generate  $\{x_k\}_{k=0}^{\infty}$  by

$$x_{k+1}=g(x_k), \quad k\geq 0.$$

 $\{x_k\}$  may not converge, e.g., g(x)=3x. However, when the sequence converges, say,

$$\lim_{k\to\infty}x_k=x^*,$$

then, since g is continuous,

$$g(x^*) = g(\lim_{k\to\infty} x_k) = \lim_{k\to\infty} g(x_k) = \lim_{k\to\infty} x_{k+1} = x^*.$$

That is,  $x^*$  is a fixed point of g.



#### Fixed-point iteration

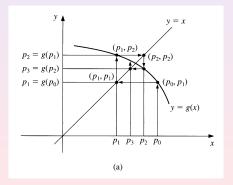
Given  $x_0$ , tolerance TOL, maximum number of iteration M.

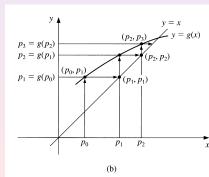
Set i = 1 and  $x = g(x_0)$ .

While  $i \leq M$  and  $|x - x_0| \geq TOL$ 

Set i = i + 1,  $x_0 = x$  and  $x = g(x_0)$ .

End While





#### Example

The equation

$$x^3 + 4x^2 - 10 = 0$$

has a unique root in [1,2]. Change the equation to the fixed-point form x = g(x).

(a) 
$$x = g_1(x) \equiv x - f(x) = x - x^3 - 4x^2 + 10$$

(b) 
$$x = g_2(x) = \left(\frac{10}{x} - 4x\right)^{1/2}$$

$$x^{3} = 10 - 4x^{2} \implies x^{2} = \frac{10}{x} - 4x \implies x = \pm \left(\frac{10}{x} - 4x\right)^{1/2}$$



(c) 
$$x = g_3(x) = \frac{1}{2} (10 - x^3)^{1/2}$$

$$4x^2 = 10 - x^3$$
  $\Rightarrow$   $x = \pm \frac{1}{2} (10 - x^3)^{1/2}$ 

(d) 
$$x = g_4(x) = \left(\frac{10}{4+x}\right)^{1/2}$$

$$x^2(x+4) = 10$$
  $\Rightarrow$   $x = \pm \left(\frac{10}{4+x}\right)^{1/2}$ 

(e) 
$$x = g_5(x) = x - \frac{x^3 + 4x^2 - 10}{3x^2 + 8x}$$

$$x = g_5(x) \equiv x - \frac{f(x)}{f'(x)}$$





# Results of the fixed-point iteration with initial point $x_0 = 1.5$

n	(a)	(b)	(c)	(d)	(e)
0	1.5	1.5	1.5	1.5	1.5
1	-0.875	0.8165	1.286953768	1.348399725	1.373333333
2	6.732	2.9969	1.402540804	1.367376372	1.365262015
3	-469.7	$(-8.65)^{1/2}$	1.345458374	1.364957015	1.365230014
4	$1.03 \times 10^{8}$		1.375170253	1.365264748	1.365230013
5			1.360094193	1.365225594	
6			1.367846968	1.365230576	
7			1.363887004	1.365229942	
8			1.365916734	1.365230022	
9			1.364878217	1.365230012	
10			1.365410062	1.365230014	
15			1.365223680	1.365230013	
20			1.365230236		
25			1.365230006		
30			1.365230013		



# Theorem (Fixed-point Theorem)

Let  $g \in C[a, b]$  be such that  $g(x) \in [a, b]$  for all  $x \in [a, b]$ . Suppose that g' exists on (a, b) and that  $\exists k \text{ with } 0 < k < 1 \text{ such that}$ 

$$|g'(x)| \le k, \ \forall \ x \in (a,b).$$

Then, for any number  $x_0$  in [a, b],

$$x_n=g(x_{n-1}), \ n\geq 1,$$

converges to the unique fixed point x in [a, b].



*Proof:* By the assumptions, a unique fixed point exists in [a,b]. Since  $g([a,b])\subseteq [a,b],\ \{x_n\}_{n=0}^\infty$  is defined and  $x_n\in [a,b]$  for all  $n\geq 0$ . Using the Mean Values Theorem and the fact that  $|g'(x)|\leq k$ , we have

$$|x-x_n|=|g(x_{n-1})-g(x)|=|g'(\xi_n)||x-x_{n-1}|\leq k|x-x_{n-1}|,$$

where  $\xi_n \in (a, b)$ . It follows that

$$|x_n - x| \le k|x_{n-1} - x| \le k^2|x_{n-2} - x| \le \dots \le k^n|x_0 - x|.$$
 (1)

Since 0 < k < 1, we have

$$\lim_{n\to\infty} k^n = 0$$

and

$$\lim_{n\to\infty} |x_n-x| \le \lim_{n\to\infty} k^n |x_0-x| = 0.$$

Hence,  $\{x_n\}_{n=0}^{\infty}$  converges to x.



### Corollary

If g satisfies the hypotheses of above theorem, then

$$|x-x_n| \le k^n \max\{x_0-a,b-x_0\}$$

and

$$|x_n - x| \le \frac{k^n}{1 - k} |x_1 - x_0|, \ \forall \ n \ge 1.$$

Proof: From (1),

$$|x_n - x| \le k^n |x_0 - x| \le k^n \max\{x_0 - a, b - x_0\}.$$

For  $n \ge 1$ , using the Mean Values Theorem,

$$|x_{n+1}-x_n|=|g(x_n)-g(x_{n-1})|\leq k|x_n-x_{n-1}|\leq \cdots \leq k^n|x_1-x_0|.$$



Thus, for  $m > n \ge 1$ ,

$$|x_{m}-x_{n}| = |x_{m}-x_{m-1}+x_{m-1}-\cdots+x_{n+1}-x_{n}|$$

$$\leq |x_{m}-x_{m-1}|+|x_{m-1}-x_{m-2}|+\cdots+|x_{n+1}-x_{n}|$$

$$\leq k^{m-1}|x_{1}-x_{0}|+k^{m-2}|x_{1}-x_{0}|+\cdots+k^{n}|x_{1}-x_{0}|$$

$$= k^{n}|x_{1}-x_{0}|\left(1+k+k^{2}+\cdots+k^{m-n-1}\right).$$

It implies that

$$|x - x_n| = \lim_{m \to \infty} |x_m - x_n| \le \lim_{m \to \infty} k^n |x_1 - x_0| \sum_{j=0}^{m-n-1} k^j$$

$$\le k^n |x_1 - x_0| \sum_{j=0}^{\infty} k^j = \frac{k^n}{1 - k} |x_1 - x_0|.$$





# Example

For previous example,  $f(x) = x^3 + 4x^2 - 10 = 0$ .

Let 
$$g_1(x) = x - x^3 - 4x^2 + 10$$
, we have

$$g_1(1) = 6$$
 and  $g_1(2) = -12$ ,

so  $g_1([1,2]) \nsubseteq [1,2]$ . Moreover,

$$g_1'(x) = 1 - 3x^2 - 8x \quad \Rightarrow \quad |g_1'(x)| \ge 1 \ \forall \ x \in [1, 2]$$

Convergence is NOT guaranteed. In fact, it almost for sure will not converge since when  $x_n$  is close to the solution  $x^*$ ,

$$|x_n - x^*| = |g_1(x_{n-1}) - g_1(x^*)| = |g_1'(c)(x_{n-1} - x^*)| > |x_{n-1} - x^*|.$$

The error is amplified whenever  $x_n$  is close to convergence. The only possibility for convergence is when  $x_n$  is far from  $x^*$  and (by chance, and very unlikely) that  $g_1(x_n) = x^*$ .

For 
$$g_3(x) = \frac{1}{2}(10 - x^3)^{1/2}$$
,  $\forall x \in [1, 1.5]$ ,

$$g_3'(x) = -\frac{3}{4}x^2(10-x^3)^{-1/2} < 0, \ \forall \ x \in [1, 1.5],$$

so  $g_3$  is strictly decreasing on [1, 1.5] and

$$1 < 1.28 \approx g_3(1.5) \le g_3(x) \le g_3(1) = 1.5, \ \forall \ x \in [1, 1.5].$$

On the other hand,

$$|g_3'(x)| \le |g_3'(1.5)| \approx 0.66, \ \forall \ x \in [1, 1.5]$$

Hence, the sequence is convergent to the fixed point.



For  $g_4(x) = \sqrt{10/(4+x)}$ , we have

$$\sqrt{\frac{10}{6}} \le g_4(x) \le \sqrt{\frac{10}{5}}, \ \forall \ x \in [1,2] \quad \Rightarrow \quad g_4([1,2]) \subseteq [1,2]$$

Moreover,

$$|g_4'(x)| = \left| \frac{-5}{\sqrt{10}(4+x)^{3/2}} \right| \le \frac{5}{\sqrt{10}(5)^{3/2}} < 0.15, \ \forall \ x \in [1,2].$$

The bound of  $|g'_4(x)|$  is much smaller than the bound of  $|g'_3(x)|$ , which explains the more rapid convergence using  $g_4$ .



#### Newton's method

Suppose that  $f: \mathbb{R} \to \mathbb{R}$  and  $f \in C^2[a, b]$ , i.e., f'' exists and is continuous. If  $f(x^*) = 0$  and  $x^* = x + h$  where h is small, then by Taylor's theorem

$$0 = f(x^*) = f(x+h)$$

$$= f(x) + f'(x)h + \frac{1}{2}f''(x)h^2 + \frac{1}{3!}f'''(x)h^3 + \cdots$$

$$= f(x) + f'(x)h + O(h^2).$$

Since h is small,  $O(h^2)$  is negligible. It is reasonable to drop  $O(h^2)$  terms. This implies

$$f(x) + f'(x)h \approx 0$$
 and  $h \approx -\frac{f(x)}{f'(x)}$ , if  $f'(x) \neq 0$ .

Hence

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



is a better approximation to  $x^*$ .

This sets the stage for the Newton-Raphson's method, which starts with an initial approximation  $x_0$  and generates the sequence  $\{x_n\}_{n=0}^{\infty}$  defined by

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

Since the Taylor's expansion of f(x) at  $x_k$  is given by

$$f(x) = f(x_k) + f'(x_k)(x - x_k) + \frac{1}{2}f''(x_k)(x - x_k)^2 + \cdots$$

At  $x_k$ , one uses the tangent line

$$y = \ell(x) = f(x_k) + f'(x_k)(x - x_k)$$

to approximate the curve of f(x) and uses the zero of the tangent line to approximate the zero of f(x).

#### Newton's Method

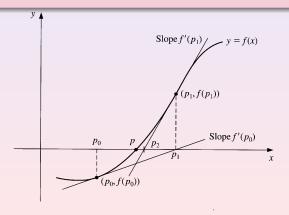
Given  $x_0$ , tolerance TOL, maximum number of iteration M.

Set i = 1 and  $x = x_0 - f(x_0)/f'(x_0)$ .

While  $i \leq M$  and  $|x - x_0| \geq TOL$ 

Set i = i + 1,  $x_0 = x$  and  $x = x_0 - f(x_0)/f'(x_0)$ .

End While





# Three stopping-technique inequalities

(a). 
$$|x_n-x_{n-1}|<\varepsilon$$
,

(b). 
$$\frac{|x_n-x_{n-1}|}{|x_n|}<\varepsilon, \quad x_n\neq 0,$$

(c).  $|f(x_n)| < \varepsilon$ .

Note that Newton's method for solving f(x) = 0

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}, \text{ for } n \ge 1$$

is just a special case of functional iteration in which

$$g(x) = x - \frac{f(x)}{f'(x)}.$$



# Example

The following table shows the convergence behavior of Newton's method applied to solving  $f(x) = x^2 - 1 = 0$ . Observe the quadratic convergence rate.

n	X <sub>n</sub>	$ e_n  \equiv  1 - x_n $
0	2.0	1
1	1.25	0.25
2	1.025	2.5e-2
3	1.0003048780488	3.048780488e-4
4	1.0000000464611	4.64611e-8
5	1.0	0





#### **Theorem**

Assume  $f(x^*) = 0$ ,  $f'(x^*) \neq 0$  and f(x), f'(x) and f''(x) are continuous on  $N_{\varepsilon}(x^*)$ . Then if  $x_0$  is chosen sufficiently close to  $x^*$ , then

$$\left\{x_{n+1}=x_n-\frac{f(x_n)}{f'(x_n)}\right\}\to x^*.$$

Proof: Define

$$g(x) = x - \frac{f(x)}{f'(x)}.$$

Find an interval  $[x^* - \delta, x^* + \delta]$  such that

$$g([x^* - \delta, x^* + \delta]) \subseteq [x^* - \delta, x^* + \delta]$$

and

$$|g'(x)| \le k < 1, \ \forall \ x \in (x^* - \delta, x^* + \delta).$$



Since f' is continuous and  $f'(x^*) \neq 0$ , it implies that  $\exists \ \delta_1 > 0$  such that  $f'(x) \neq 0 \ \forall \ x \in [x^* - \delta_1, x^* + \delta_1] \subseteq [a, b]$ . Thus, g is defined and continuous on  $[x^* - \delta_1, x^* + \delta_1]$ . Also

$$g'(x) = 1 - \frac{f'(x)f'(x) - f(x)f''(x)}{[f'(x)]^2} = \frac{f(x)f''(x)}{[f'(x)]^2},$$

for  $x \in [x^* - \delta_1, x^* + \delta_1]$ . Since f'' is continuous on [a, b], we have g' is continuous on  $[x^* - \delta_1, x^* + \delta_1]$ . By assumption  $f(x^*) = 0$ , so

$$g'(x^*) = \frac{f(x^*)f''(x^*)}{|f'(x^*)|^2} = 0.$$

Since g' is continuous on  $[x^* - \delta_1, x^* + \delta_1]$  and  $g'(x^*) = 0$ ,  $\exists \ \delta$  with  $0 < \delta < \delta_1$  and  $k \in (0,1)$  such that

$$|g'(x)| \le k, \ \forall \ x \in [x^* - \delta, x^* + \delta].$$



Claim:  $g([x^* - \delta, x^* + \delta]) \subseteq [x^* - \delta, x^* + \delta]$ . If  $x \in [x^* - \delta, x^* + \delta]$ , then, by the Mean Value Theorem,  $\exists \xi$  between x and  $x^*$  such that

$$|g(x) - g(x^*)| = |g'(\xi)||x - x^*|.$$

It implies that

$$|g(x) - x^*| = |g(x) - g(x^*)| = |g'(\xi)||x - x^*|$$
  
 
$$\leq k|x - x^*| < |x - x^*| < \delta.$$

Hence,  $g([x^* - \delta, x^* + \delta]) \subseteq [x^* - \delta, x^* + \delta]$ . By the Fixed-Point Theorem, the sequence  $\{x_n\}_{n=0}^{\infty}$  defined by

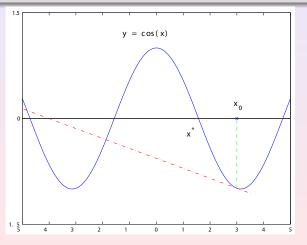
$$x_n = g(x_{n-1}) = x_{n-1} - \frac{f(x_{n-1})}{f'(x_{n-1})}, \text{ for } n \ge 1,$$

converges to  $x^*$  for any  $x_0 \in [x^* - \delta, x^* + \delta]$ .



## Example

When Newton's method applied to  $f(x) = \cos x$  with starting point  $x_0 = 3$ , which is close to the root  $\frac{\pi}{2}$  of f, it produces  $x_1 = -4.01525, x_2 = -4.8526, \cdots$ , which converges to another root  $-\frac{3\pi}{2}$ .





## Secant method

# Disadvantage of Newton's method

In many applications, the derivative f'(x) is very expensive to compute, or the function f(x) is not given in an algebraic formula so that f'(x) is not available.

By definition,

$$f'(x_{n-1}) = \lim_{x \to x_{n-1}} \frac{f(x) - f(x_{n-1})}{x - x_{n-1}}.$$

Letting  $x = x_{n-2}$ , we have

$$f'(x_{n-1}) \approx \frac{f(x_{n-2}) - f(x_{n-1})}{x_{n-2} - x_{n-1}} = \frac{f(x_{n-1}) - f(x_{n-2})}{x_{n-1} - x_{n-2}}.$$

Using this approximation for  $f'(x_{n-1})$  in Newton's formula gives

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



From geometric point of view, we use a secant line through  $x_{n-1}$  and  $x_{n-2}$  instead of the tangent line to approximate the function at the point  $x_{n-1}$ . The slope of the secant line is

$$s_{n-1} = \frac{f(x_{n-1}) - f(x_{n-2})}{x_{n-1} - x_{n-2}}$$

and the equation is

$$M(x) = f(x_{n-1}) + s_{n-1}(x - x_{n-1}).$$

The zero of the secant line

$$x = x_{n-1} - \frac{f(x_{n-1})}{s_{n-1}} = x_{n-1} - f(x_{n-1}) \frac{x_{n-1} - x_{n-2}}{f(x_{n-1}) - f(x_{n-2})}$$

is then used as a new approximate  $x_n$ .



#### Secant Method

Given  $x_0, x_1$ , tolerance TOL, maximum number of iteration M.

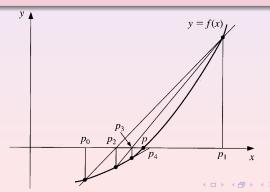
Set 
$$i = 2$$
;  $y_0 = f(x_0)$ ;  $y_1 = f(x_1)$ ;

$$x = x_1 - y_1(x_1 - x_0)/(y_1 - y_0).$$

While 
$$i \leq M$$
 and  $|x - x_1| \geq TOL$ 

Set 
$$i = i + 1$$
;  $x_0 = x_1$ ;  $y_0 = y_1$ ;  $x_1 = x$ ;  $y_1 = f(x)$ ;  $x = x_1 - y_1(x_1 - x_0)/(y_1 - y_0)$ .

End While





#### Method of False Position

- **1** Choose initial approximations  $x_0$  and  $x_1$  with  $f(x_0)f(x_1) < 0$ .
- 2  $x_2 = x_1 f(x_1)(x_1 x_0)/(f(x_1) f(x_0))$
- ① Decide which secant line to use to compute  $x_3$ : If  $f(x_2)f(x_1) < 0$ , then  $x_1$  and  $x_2$  bracket a root, i.e.,

$$x_3 = x_2 - f(x_2)(x_2 - x_1)/(f(x_2) - f(x_1))$$

Else,  $x_0$  and  $x_2$  bracket a root, i.e.,

$$x_3 = x_2 - f(x_2)(x_2 - x_0)/(f(x_2) - f(x_0))$$

End if



### Method of False Position

Given  $x_0, x_1$ , tolerance TOL, maximum number of iteration M.

Set 
$$i = 2$$
;  $y_0 = f(x_0)$ ;  $y_1 = f(x_1)$ ;  $x = x_1 - y_1(x_1 - x_0)/(y_1 - y_0)$ .

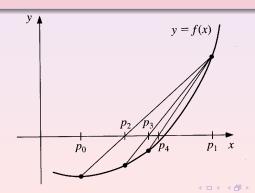
While 
$$i \leq M$$
 and  $|x - x_1| \geq TOL$ 

Set 
$$i = i + 1$$
;  $y = f(x)$ .

If 
$$y \cdot y_1 < 0$$
, then set  $x_0 = x_1$ ;  $y_0 = y_1$ .

Set 
$$x_1 = x$$
;  $y_1 = y$ ;  $x = x_1 - y_1(x_1 - x_0)/(y_1 - y_0)$ .

End While





# Error analysis for iterative methods

### **Definition**

Let  $\{x_n\} \to x^*$ . If there are positive constants c and  $\alpha$  such that

$$\lim_{n\to\infty}\frac{|x_{n+1}-x^*|}{|x_n-x^*|^{\alpha}}=c,$$

then we say the rate of convergence is of order  $\alpha$ .

We say that the rate of convergence is

- 1 linear if  $\alpha = 1$  and 0 < c < 1.
- 2 superlinear if

$$\lim_{n \to \infty} \frac{|x_{n+1} - x^*|}{|x_n - x^*|} = 0;$$

3 quadratic if  $\alpha = 2$ .

Suppose that  $\{x_n\}_{n=0}^{\infty}$  and  $\{\tilde{x}_n\}_{n=0}^{\infty}$  are linearly and quadratically convergent to  $x^*$ , respectively, with the same constant c=0.5. For simplicity, suppose that

$$\frac{|x_{n+1}-x^*|}{|x_n-x^*|}\approx c \quad \text{and} \quad \frac{|\tilde{x}_{n+1}-x^*|}{|\tilde{x}_n-x^*|^2}\approx c.$$

These imply that

$$|x_n - x^*| \approx c|x_{n-1} - x^*| \approx c^2|x_{n-2} - x^*| \approx \cdots \approx c^n|x_0 - x^*|,$$

and

$$|\tilde{x}_{n} - x^{*}| \approx c|\tilde{x}_{n-1} - x^{*}|^{2} \approx c \left[c|\tilde{x}_{n-2} - x^{*}|^{2}\right]^{2} = c^{3}|\tilde{x}_{n-2} - x^{*}|^{4}$$

$$\approx c^{3} \left[c|\tilde{x}_{n-3} - x^{*}|^{2}\right]^{4} = c^{7}|\tilde{x}_{n-3} - x^{*}|^{8}$$

$$\approx \cdots \approx c^{2^{n}-1}|\tilde{x}_{n} - x^{*}|^{2^{n}}.$$



### Remark

Quadratically convergent sequences generally converge much more quickly than those that converge only linearly.

### **Theorem**

Let  $g \in C[a, b]$  with  $g([a, b]) \subseteq [a, b]$ . Suppose that g' is continuous on (a, b) and  $\exists k \in (0, 1)$  such that

$$|g'(x)| \le k, \ \forall \ x \in (a,b).$$

If  $g'(x^*) \neq 0$ , then for any  $x_0 \in [a, b]$ , the sequence

$$x_n = g(x_{n-1}), \text{ for } n \ge 1$$

converges only linearly to the unique fixed point  $x^*$  in [a, b].



### Proof:

- By the Fixed-Point Theorem, the sequence  $\{x_n\}_{n=0}^{\infty}$  converges to  $x^*$ .
- Since g' exists on (a,b), by the Mean Value Theorem,  $\exists \ \xi_n$  between  $x_n$  and  $x^*$  such that

$$x_{n+1} - x^* = g(x_n) - g(x^*) = g'(\xi_n)(x_n - x^*).$$

- $\bullet :: \{x_n\}_{n=0}^{\infty} \to x^* \quad \Rightarrow \quad \{\xi_n\}_{n=0}^{\infty} \to x^*$
- Since g' is continuous on (a, b), we have

$$\lim_{n\to\infty}g'(\xi_n)=g'(x^*).$$

Thus,

$$\lim_{n \to \infty} \frac{|x_{n+1} - x^*|}{|x_n - x^*|} = \lim_{n \to \infty} |g'(\xi_n)| = |g'(x^*)|.$$

Hence, if  $g'(x^*) \neq 0$ , fixed-point iteration exhibits linear convergence.



#### Theorem

Let  $x^*$  be a fixed point of g and I be an open interval with  $x^* \in I$ . Suppose that  $g'(x^*) = 0$  and g'' is continuous with

$$|g''(x)| < M, \ \forall \ x \in I.$$

Then  $\exists \ \delta > 0$  such that

$$\{x_n = g(x_{n-1})\}_{n=1}^{\infty} \rightarrow x^* \text{ for } x_0 \in [x^* - \delta, x^* + \delta]$$

at least quadratically. Moreover,

$$|x_{n+1}-x^*|<\frac{M}{2}|x_n-x^*|^2$$
, for sufficiently large  $n$ .





### Proof:

• Since  $g'(x^*)=0$  and g' is continuous on I,  $\exists \ \delta$  such that  $[x^*-\delta,x^*+\delta]\subset I$  and

$$|g'(x)| \le k < 1, \ \forall \ x \in [x^* - \delta, x^* + \delta].$$

In the proof of the convergence for Newton's method, we have

$$\{x_n\}_{n=0}^{\infty} \subset [x^* - \delta, x^* + \delta].$$

• Consider the Taylor expansion of  $g(x_n)$  at  $x^*$ 

$$x_{n+1} = g(x_n) = g(x^*) + g'(x^*)(x_n - x^*) + \frac{g''(\xi)}{2}(x_n - x^*)^2$$
$$= x^* + \frac{g''(\xi)}{2}(x_n - x^*)^2,$$

where  $\xi$  lies between  $x_n$  and  $x^*$ .



Since

$$|g'(x)| \le k < 1, \ \forall \ x \in [x^* - \delta, x^* + \delta]$$

and

$$g([x^* - \delta, x^* + \delta]) \subseteq [x^* - \delta, x^* + \delta],$$

it follows that  $\{x_n\}_{n=0}^{\infty}$  converges to  $x^*$ .

• But  $\xi_n$  is between  $x_n$  and  $x^*$  for each n, so  $\{\xi_n\}_{n=0}^{\infty}$  also converges to  $x^*$  and

$$\lim_{n\to\infty} \frac{|x_{n+1}-x^*|}{|x_n-x^*|^2} = \frac{|g''(x^*)|}{2} < \frac{M}{2}.$$

• It implies that  $\{x_n\}_{n=0}^{\infty}$  is quadratically convergent to  $x^*$  if  $g''(x^*) \neq 0$  and

$$|x_{n+1}-x^*|<\frac{M}{2}|x_n-x^*|^2$$
, for sufficiently large  $n$ .





# Example

Recall that Newton's method  $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$  corresponds to  $g(x) = x - \frac{f(x)}{f'(x)}$ . Suppose that f(x) has a m-fold root at  $x^*$ , that is

$$f(x) = (x - x^*)^m q(x), \qquad q(x^*) \neq 0.$$

Let  $\mu(x) = \frac{f(x)}{f'(x)} = (x - x^*) \frac{g(x)}{mq(x) + (x - x^*)q'(x)}$ , it is easy to see that  $\mu'(x^*) = \frac{1}{m}$ . It follows that  $0 \le g'(x_*) = 1 - \frac{1}{m} < 1$ . Hence Newton's method is locally convergent. Moreover, it converges quadratically for simple roots (m = 1) and linearly for multiple roots (m > 1).

Remedy for slow convergence on multiple roots (m > 1):

- If m is known, take  $x_{n+1} = x_n \frac{mf(x_n)}{f'(x_n)}$ .
- If m is not known, take  $x_{n+1} = x_n \frac{\mu(x_n)}{\mu'(x_n)}$ , since  $\mu(x) = \frac{f(x)}{f'(x)} = \frac{O(x-x^*)^m}{O(x-x^*)^{m-1}} = O(x-x^*)$  always has a simple root at  $x^*$  for any m > 1. This is known as modified Newton's method.

# Global Convergence for Convex (Concave) Functions

### **Theorem**

If  $f \in C^2$ , f'' > 0 and f(x) = 0 has a root, then Newton's method always converges to a root  $x^*$  for any initial  $x_0$ .

## Proof:

It suffices to consider the case where f'>0, f''>0 and f(x)=0 has a root. In this case, the root  $x^*$  is unique. Define  $e_n=x_n-x^*$ . Since  $x_{n+1}=x_n-\frac{f(x_n)}{f'(x_n)}$ . It follows that

$$e_{n+1} = e_n - \frac{f(x_n)}{f'(x_n)}.$$
 (2)

Moreover, since  $f(x^*) = f(x_n) + f'(x_n)(x^* - x_n) + \frac{f''(\xi_n)}{2}(x^* - x_n)^2$ , we also have  $f(x_n) = f'(x_n)e_n - \frac{f''(\xi_n)}{2}e_n^2$ . Therefore

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

(3)

Consequently  $x_{n+1} > x^*$  and  $f(x_{n+1}) > 0$  for all  $n \ge 0$ 

Moreover 
$$e_{n+1} = e_n - \frac{f(x_n)}{f'(x_n)} < e_n$$
, we conclude that

$$0 < \dots < x_{n+1} < x_n < \dots < x_1$$

and  $x_n$  converges monotonically to some  $\tilde{x}$  satisfying  $\tilde{x} = \tilde{x} - \frac{f(\tilde{x})}{f'(\tilde{x})}$ , that is  $f(\tilde{x}) = 0$ , thus  $\tilde{x} = x^*$  by uniqueness of the root.

The proof for other cases

- f' < 0, f'' > 0, f(x) = 0 has a root.
- f'' > 0, has two distinct roots.
- f'' > 0, has a double root. are similar.

So is the concave case (f'' < 0).



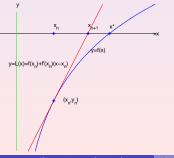
# Alternative Error Estimate for Newton's Method

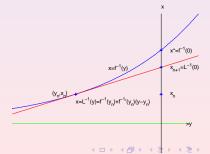
Suppose  $f'(x^*) \neq 0$ , then both f(x) and its linearization at  $(x_n, y_n)$ , L(x), are locally invertible (Inverse Function Theorem). The formula of the tangent lines are given by

$$L(x) = f(x_n) + \frac{df(x_n)}{dx}(x - x_n)$$

and

$$L^{-1}(y) = f^{-1}(y_n) + \frac{df^{-1}(y_n)}{dy}(y - y_n) = x_n + \frac{1}{f'(x_n)}(y - y_n)$$







Since  $x^* = f^{-1}(0)$  and  $x_{n+1} = L^{-1}(0)$ , the error estimate for Newton's method reduces to error estimate between  $f^{-1}(y)$  and its linearization approximation  $L^{-1}(y)$  at y = 0. From standard analysis, the error is proportional to  $(0 - y_n)^2$ :

$$|x_{n+1}-x^*|=|L^{-1}(0)-f^{-1}(0)|=\frac{1}{2}\left|\frac{d^2f^{-1}}{dy^2}(\eta_n)(y_n-0)^2\right|$$

$$= \frac{1}{2} \left| \frac{d^2 f^{-1}}{dy^2} (\eta_n) \right| \left( f(x_n) - f(x^*) \right)^2 = \left( \frac{1}{2} \left| \frac{d^2 f^{-1}}{dy^2} (\eta_n) \right| \cdot \left( f'(\xi_n) \right)^2 \right) (x_n - x^*)^2$$

The main advantage of this formulation:

Higher order approximations of  $f^{-1}(0)$ , such as quadratic approximation, gives rise to higher order iteration schemes for solving the original equation f(x) = 0.



# Error Analysis of Secant Method

Reference: D. Kincaid and W. Cheney, "Numerical analysis" Let  $x^*$  denote the exact solution of f(x) = 0,  $e_k = x_k - x^*$  be the errors at the k-th step. Then

$$e_{k+1} = x_{k+1} - x^*$$

$$= x_k - f(x_k) \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} - x^*$$

$$= \frac{1}{f(x_k) - f(x_{k-1})} [(x_{k-1} - x^*)f(x_k) - (x_k - x^*)f(x_{k-1})]$$

$$= \frac{1}{f(x_k) - f(x_{k-1})} (e_{k-1}f(x_k) - e_k f(x_{k-1}))$$

$$= e_k e_{k-1} \left( \frac{\frac{1}{e_k} f(x_k) - \frac{1}{e_{k-1}} f(x_{k-1})}{x_k - x_{k-1}} \cdot \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} \right)$$





To estimate the numerator  $\frac{\frac{1}{e_k}f(x_k)-\frac{1}{e_{k-1}}f(x_k)}{x_k-x_{k-1}}$ , we apply the Taylor's theorem

$$f(x_k) = f(x^* + e_k) = f(x^*) + f'(x^*)e_k + \frac{1}{2}f''(x^*)e_k^2 + O(e_k^3),$$

to get

$$\frac{1}{e_k}f(x_k) = f'(x^*) + \frac{1}{2}f''(x^*)e_k + O(e_k^2).$$

Similarly,

$$\frac{1}{e_{k-1}}f(x_{k-1})=f'(x^*)+\frac{1}{2}f''(x^*)e_{k-1}+O(e_{k-1}^2).$$

Hence

$$\frac{1}{e_k}f(x_k) - \frac{1}{e_{k-1}}f(x_{k-1}) \approx \frac{1}{2}(e_k - e_{k-1})f''(x^*).$$

Since  $x_k - x_{k-1} = e_k - e_{k-1}$  and

$$\frac{x_k-x_{k-1}}{f(x_k)-f(x_{k-1})}\to \frac{1}{f'(x^*)},$$



we have

$$e_{k+1} \approx e_k e_{k-1} \left( \frac{\frac{1}{2} (e_k - e_{k-1}) f''(x^*)}{e_k - e_{k-1}} \cdot \frac{1}{f'(x^*)} \right) = \frac{1}{2} \frac{f''(x^*)}{f'(x^*)} e_k e_{k-1}$$

$$\equiv C e_k e_{k-1}$$
(4)

To estimate the convergence rate, we assume

$$|e_{k+1}| \approx \eta |e_k|^{\alpha},$$

where  $\eta > 0$  and  $\alpha > 0$  are constants, i.e.,

$$rac{|e_{k+1}|}{\eta|e_k|^{lpha}} 
ightarrow 1 \quad ext{as} \quad k 
ightarrow \infty.$$

Then  $|e_k| \approx \eta |e_{k-1}|^{\alpha}$  which implies  $|e_{k-1}| \approx \eta^{-1/\alpha} |e_k|^{1/\alpha}$ . Hence (4) gives

$$\eta |e_k|^{\alpha} \approx C |e_k| \eta^{-1/\alpha} |e_k|^{1/\alpha} \implies C^{-1} \eta^{1+\frac{1}{\alpha}} \approx |e_k|^{1-\alpha+\frac{1}{\alpha}}.$$

Since  $|e_k| \to 0$  as  $k \to \infty$ , and  $C^{-1}\eta^{1+\frac{1}{\alpha}}$  is a nonzero constant,

$$1 - \alpha + \frac{1}{\alpha} = 0 \implies \alpha = \frac{1 + \sqrt{5}}{2} \approx 1.62.$$



This result implies that  $C^{-1}\eta^{1+\frac{1}{\alpha}} \to 1$  and

$$\eta \to C^{\frac{\alpha}{1+\alpha}} = \left(\frac{f''(x^*)}{2f'(x^*)}\right)^{0.62}.$$

In summary, we have shown that

$$|e_{k+1}| = \eta |e_k|^{\alpha}, \quad \alpha \approx 1.62,$$

that is, the rate of convergence is superlinear.

Rate of convergence:

• secant method: superlinear

Newton's method: quadratic

• bisection method: linear



## Each iteration of method requires

- secant method: one function evaluation
- Newton's method: two function evaluation, namely,  $f(x_k)$  and  $f'(x_k)$ .  $\Rightarrow$  two steps of secant method are comparable to one step of Newton's method. Thus

$$|e_{k+2}| \approx \eta |e_{k+1}|^{\alpha} \approx \eta^{1+\alpha} |e_k|^{\frac{3+\sqrt{5}}{2}} \approx \eta^{1+\alpha} |e_k|^{2.62}.$$

⇒ secant method is more efficient than Newton's method.

### Remark

Two steps of secant method would require a little more work than one step of Newton's method.



# Accelerating convergence

## Aitken's $\Delta^2$ method

- Accelerate the convergence of a sequence that is linearly convergent.
- Suppose  $\{x_n\}_{n=0}^{\infty}$  is a linearly convergent sequence with limit y. Construct  $\{\hat{x}_n\}_{n=0}^{\infty}$  that converges more rapidly to x than  $\{x_n\}_{n=0}^{\infty}$ .

For *n* sufficiently large,

$$\frac{x_{n+1}-x}{x_n-x}\approx\frac{x_{n+2}-x}{x_{n+1}-x}.$$

Then

$$(x_{n+1}-x)^2 \approx (x_{n+2}-x)(x_n-x),$$

SO

$$x_{n+1}^2 - 2x_{n+1}x + x^2 \approx x_{n+2}x_n - (x_{n+2} + x_n)x + x^2$$



and

$$(x_{n+2} + x_n - 2x_{n+1})x \approx x_{n+2}x_n - x_{n+1}^2$$

Solving for x gives

$$x \approx \frac{x_{n+2}x_n - x_{n+1}^2}{x_{n+2} - 2x_{n+1} + x_n}$$

$$= \frac{x_nx_{n+2} - 2x_nx_{n+1} + x_n^2 - x_n^2 + 2x_nx_{n+1} - x_{n+1}^2}{x_{n+2} - 2x_{n+1} + x_n}$$

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

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

## Aitken's $\Delta^2$ method

$$\hat{x}_n = x_n - \frac{(x_{n+1} - x_n)^2}{(x_{n+2} - x_{n+1}) - (x_{n+1} - x_n)} := \{\Delta^2\} x_n.$$
 (5)

#### **Theorem**

Suppose  $\{x_n\}_{n=0}^{\infty} \to x$  linearly and

$$\lim_{n\to\infty}\frac{x_{n+1}-x}{x_n-x}<1.$$

Then  $\{\hat{x}_n\}_{n=0}^{\infty} \to x$  faster than  $\{x_n\}_{n=0}^{\infty}$  in the sense that

$$\lim_{n\to\infty}\frac{\hat{x}_n-x}{x_n-x}=0.$$

• Aitken's  $\Delta^2$  method constructs the terms in order:

$$x_1 = g(x_0), \quad x_2 = g(x_1), \quad \hat{x}_0 = \{\Delta^2\}(x_0), \quad x_3 = g(x_2), \\ \hat{x}_1 = \{\Delta^2\}(x_1), \quad \hat{x}_2 = \{\Delta^2\}(x_2), \quad \cdots,$$

This is based on the assumption that  $|\hat{x}_0 - x| < |x_2 - x|$ ,  $|\hat{x}_1 - x| < |x_3 - x|$ , etc.



## Example

The sequence  $\{x_n = \cos(\frac{1}{n})\}_{n=1}^{\infty}$  converges linearly to x = 1.

n	X <sub>n</sub>	$e_n$	$\hat{x}_n$	ê <sub>n</sub>
1	0.54030	0.45969	0.96178	0.03822
2	0.87758	0.12242	0.98213	0.01787
3	0.94496	0.05504	0.98979	0.01021
4	0.96891	0.03109	0.99342	0.00658
5	0.98007	0.01993	0.99541	0.00459
6	0.98614	0.01386		
7	0.98981	0.01019		

• Note that  $\lim_{n\to\infty}\frac{x_{n+1}-x}{x_n-x}=1$ . The assumption in previous Theorem is not satisfied. In this case,  $\{\hat{x}_n\}_{n=1}^\infty$  converges more rapidly to x=1 than  $\{x_{n+2}\}_{n=1}^\infty$ , but is of the same order. In fact  $\hat{e}_n/e_{n+2}\sim 1/3$  for large n. (Why?)

Steffensen's method constructs the terms in order:

$$x_0^{(0)} = x_0, x_1^{(0)} (= x_1) = g(x_0^{(0)}), x_2^{(0)} (= x_2) = g(x_1^{(0)}),$$

$$x_0^{(1)} (= x_3) = \{\Delta^2\}(x_0^{(0)}), x_1^{(1)} (= x_4) = g(x_0^{(1)}), x_2^{(1)} (= x_5) = g(x_1^{(1)}),$$

$$x_0^{(2)} (= x_6) = \{\Delta^2\}(x_0^{(0)}), x_1^{(2)} (= x_7) = g(x_0^{(2)}), x_2^{(2)} (= x_8) = g(x_1^{(2)}),$$

$$\cdots,$$

# Steffensen's method (To find a solution of x = g(x))

Given  $x_0$ , tolerance TOL, maximum number of iteration M.

Set 
$$i = 1$$
.

While 
$$i < M$$

Set 
$$x_1 = g(x_0)$$
;  $x_2 = g(x_1)$ ;  $x = x_0 - (x_1 - x_0)^2 / (x_2 - 2x_1 + x_0)$ .

If 
$$|x - x_0| < ToI$$
, then STOP.

Set 
$$i = i + 1$$
;  $x_0 = x$ .

End While

40 40 40 40 40 600

### **Theorem**

Suppose that x = g(x) has the solution  $x^*$  with  $g'(x^*) \neq 1$ . If  $\exists \ \delta > 0$  such that  $g \in C^3[x^* - \delta, x^* + \delta]$ , then Steffensen's method gives quadratic convergence for any  $x_0 \in [x^* - \delta, x^* + \delta]$ .

## Proof:

We denote by  $x_0$ ,  $x_1$ ,  $x_2$ ,  $\cdots$ , (instead of  $x_0^{(i)}$ ,  $x_1^{(i)}$ ,  $x_2^{(i)}$ ), the sequence generated by Steffensen's method.

We will show that  $|x_3-x| \le C|x_0-x|^2$ ,  $|x_6-x| \le C|x_3-x|^2$ , etc. to establish quadratic convergence. Denote by  $\Delta_i = x_i - x^*$ , we have

$$\Delta_{1} = x_{1} - x^{*} = g(x_{0}) - g(x^{*})$$

$$= g'(x^{*})(x_{0} - x^{*}) + \frac{g''(x^{*})}{2}(x_{0} - x^{*})^{2} + O(\Delta_{0}^{3})$$

$$\Delta_{2} = x_{2} - x^{*} = g(x_{1}) - g(x^{*})$$

$$= g'(x^{*})(x_{1} - x^{*}) + \frac{g''(x^{*})}{2}(x_{1} - x^{*})^{2} + O(\Delta_{1}^{3})$$

$$= g'(x^{*})^{2}\Delta_{0} + (\frac{g'(x^{*})g''(x^{*})}{2} + \frac{g''(x^{*})g'(x^{*})^{2}}{2})\Delta_{0}^{2} + O(\Delta_{0}^{3})$$

$$\begin{split} x_3 &= x_0 - \frac{(x_1 - x_0)^2}{x_0 - 2x_1 + x_2} \\ \Delta_3 &= \Delta_0 - \frac{(\Delta_1 - \Delta_0)^2}{\Delta_0 - 2\Delta_1 + \Delta_2} \\ &= \Delta_0 - \frac{\left((g'(x) - 1)\Delta_0 + \frac{g''(x)}{2}\Delta_0^2 + O(\Delta_0^3)\right)^2}{\left(g'^2(x) - 2g'(x) + 1\right)\Delta_0 + \frac{g''(x)}{2}\left(g'^2(x) + g'(x) - 2\right)\Delta_0^2 + O(\Delta_0^3)} \\ &= \Delta_0 - \Delta_0 \left(\frac{(g'(x) - 1)^2 + g''(x)(g'(x) - 1)\Delta_0 + O(\Delta_0^2)}{\left(g'(x) - 1\right)^2 + \frac{g''(x)}{2}\left(g'(x) + 2\right)\left(g'(x) - 1\right)\Delta_0 + O(\Delta_0^2)}\right) \\ &= \Delta_0 - \Delta_0 \left(1 - \frac{g'(x^*)g''(x^*)}{2(g'(x^*) - 1)}\Delta_0 + O(\Delta_0^2)\right), \quad \text{if } g'(x^*) \neq 1 \end{split}$$

It follows that  $x_3 - x \approx C(x_0 - x)^2$ ,  $x_6 - x \approx C(x_3 - x)^2$ , etc. with  $C = \frac{g'(x^*)g''(x^*)}{2(g'(x^*)-1)}$  if  $g'(x^*) \neq 1$ .



# Zeros of polynomials and Müller's method

• Horner's method:

Goal: Find successively all roots of a polynomial

$$P(x) = a_0 + a_1 x + a_2 x^2 + \dots + a_{n-1} x^{n-1} + a_n x^n$$
 (6)

with minimal computational cost.

The key step is to efficiently compute the quotient  $P(x)/(x-x^*)$  when a root  $x^*$  of P(x) has been found (eg. by Newton's method), or more generally, to find the quotient Q(x) and the remainder  $b_0$  such that

$$P(x) = (x - x_k)Q(x) + b_0,$$
 (7)

for any given  $x_k$ . As a byproduct, one obtains  $P'(x_k) = Q(x_k)$  from (7) which can be utilized in the Newton-Raphson iteration  $x_{k+1} = x_k - \frac{P(x_k)}{P'(x_k)}$ . The coefficients of Q(x) can be obtained by assuming

$$Q(x) = b_1 + b_2 x + \cdots + b_n x^{n-1}$$

and then comparing the coefficients in (6) and (7)

We have

$$b_0 + (x - x_k)Q(x) = b_0 + (x - x_k) (b_1 + b_2x + \dots + b_nx^{n-1})$$

$$= (b_0 - b_1x_k) + (b_1 - b_2x_k)x + \dots + (b_{n-1} - b_nx_k)x^{n-1} + b_nx^n$$

$$= a_0 + a_1x + \dots + a_nx^n = P(x).$$

and therefore

$$b_n = a_n,$$
  
 $b_j = a_j + b_{j+1}x_k,$  for  $j = n - 1, n - 2, \dots, 1, 0,$ 

Moreover, the evaluation of  $Q(x_k)$  can be obtained through the nested expression:

$$Q(x) = b_1 + x (b_2 + x (b_3 + \cdots + x (b_{n-1} + xb_n)))$$

that is, let  $c_n = b_n (= a_n)$ , and for  $j = n - 1, n - 2, \dots, 1$ ,

$$c_j = b_j + c_{j+1}x_k,$$



then  $Q(x_k) = c_1$ .

# Horner's method (Evaluate $P(x_k)$ and $P'(x_k) = Q(x_k)$ )

Set 
$$y = a_n$$
;  $z = a_n$   $(b_n = a_n)$ ;  $c_n = a_n$ .  
For  $j = n - 1$ ,  $n - 2$ , ...,  $1$   
Set  $y = a_j + yx_k$ ;  $z = y + zx_k$   $(b_j = a_j + b_{j+1}x_k)$ ;  $c_j = b_j + c_{j+1}x_k$ .  
End for  
Set  $y = a_0 + yx_k$   $(b_0 = a_0 + b_1x_k)$ .  
Output  $P(x_k) = y$   $(= b_0)$ ;  $P'(x_k) = z$   $(= c_1)$ .

If  $x_N$  is an approximate zero of P, then

$$P(x) = (x - x_N)Q(x) + b_0 = (x - x_N)Q(x) + P(x_N) \approx (x - x_N)Q(x) \equiv (x - \hat{x}_1)Q_1(x).$$

So  $x - \hat{x}_1$  is an approximate factor of P(x) and we can find a second approximate zero of P by applying Newton's method to  $Q_1(x)$ . The procedure is called deflation.



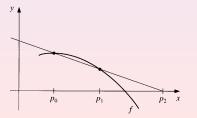
• Müller's method: Find complex roots of a polynomial P(x) (or any complex valued function  $f: \mathbb{C} \mapsto \mathbb{C}$ ):

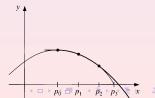
### Theorem

If z = a + ib is a complex zero of multiplicity m of P(x) with real coefficients, then  $\bar{z} = a - bi$  is also a zero of multiplicity m of P(x) and  $(x^2 - 2ax + a^2 + b^2)^m$  is a factor of P(x).

Secant method: Given  $p_0$  and  $p_1$ , determine  $p_2$  as the intersection of the x-axis with the line through  $(p_0, f(p_0))$  and  $(p_1, f(p_1))$ .

Müller's method: Given  $p_0$ ,  $p_1$  and  $p_2$ , determine  $p_3$  by the intersection of the x-axis with the parabola through  $(p_0, f(p_0))$ ,  $(p_1, f(p_1))$  and  $(p_2, f(p_2))$ .







Let

$$P(x) = a(x - p_2)^2 + b(x - p_2) + c$$

that passes through  $(p_0, f(p_0))$ ,  $(p_1, f(p_1))$  and  $(p_2, f(p_2))$ . Then

$$f(p_0) = a(p_0 - p_2)^2 + b(p_0 - p_2) + c,$$
  

$$f(p_1) = a(p_1 - p_2)^2 + b(p_1 - p_2) + c,$$
  

$$f(p_2) = a(p_2 - p_2)^2 + b(p_2 - p_2) + c = c.$$

It follows that

$$c = f(p_2),$$

$$b = \frac{(p_0 - p_2)^2 [f(p_1) - f(p_2)] - (p_1 - p_2)^2 [f(p_0) - f(p_2)]}{(p_0 - p_2)(p_1 - p_2)(p_0 - p_1)},$$

$$a = \frac{(p_1 - p_2) [f(p_0) - f(p_2)] - (p_0 - p_2) [f(p_1) - f(p_2)]}{(p_0 - p_2)(p_1 - p_2)(p_0 - p_1)}.$$



To determine  $p_3$ , a zero of P, we apply the quadratic formula to P(x) = 0 and get

$$p_3 - p_2 = \frac{2c}{b \pm \sqrt{b^2 - 4ac}}. (8)$$

If a, b, c are all real, we can choose

$$p_3 = p_2 + \frac{2c}{b + sgn(b)\sqrt{b^2 - 4ac}}$$

such that the denominator will be largest in magnitude. The selected  $p_3$  is the one closer to  $p_2$  among those given in (8).

In case a, b, c are complex, the selection principle for  $p_3$  can be modified accordingly.



# Müller's method (Find a solution of f(x) = 0)

Given  $p_0, p_1, p_2$ ; tolerance TOL; maximum number of iterations M Set  $h_1 = p_1 - p_0$ ;  $h_2 = p_2 - p_1$ ;  $\delta_1 = (f(p_1) - f(p_0))/h_1; \ \delta_2 = (f(p_2) - f(p_1))/h_2;$  $d = (\delta_2 - \delta_1)/(h_2 + h_1)$ ; i = 3. While i < MSet  $b = \delta_2 + h_2 d$ ;  $D = \sqrt{b^2 - 4f(p_2)d}$ . If |b-D| < |b+D|, then set E = b+D else set E = b-D. Set  $h = -2f(p_2)/E$ ;  $p = p_2 + h$ . If |h| < TOL, then STOP. Set  $p_0 = p_1$ ;  $p_1 = p_2$ ;  $p_2 = p$ ;  $h_1 = p_1 - p_0$ ;  $h_2 = p_2 - p_1$ ;  $\delta_1 = (f(p_1) - f(p_0))/h_1; \ \delta_2 = (f(p_2) - f(p_1))/h_2;$  $d = (\delta_2 - \delta_1)/(h_2 + h_1); i = i + 1.$ End while