

Solutions of Equations in One Variable

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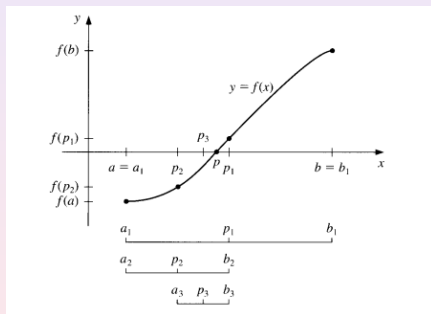
Outline



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 δ and ε , this algorithm tries to locate one root of $f(x)$.

Compute $u = f(a)$, $v = f(b)$, and $e = b - a$

If $\text{sign}(u) = \text{sign}(v)$, **then** stop

For $k = 1, 2, \dots, M$

$e = e/2$, $c = a + e$, $w = f(c)$

If $|e| < \delta$ or $|w| < \varepsilon$, **then** stop

If $\text{sign}(w) \neq \text{sign}(u)$

$b = c$, $v = w$

Else

$a = c$, $u = w$

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
- 3 $|f(c_k)| < \varepsilon$.

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

$$\begin{aligned} a &= a_0 \leq a_1 \leq a_2 \leq \dots \leq b_0 = b \\ \Rightarrow \quad &\{a_n\} \text{ and } \{b_n\} \text{ are bounded} \\ \Rightarrow \quad &\lim_{n \rightarrow \infty} a_n \text{ and } \lim_{n \rightarrow \infty} b_n \text{ exist} \end{aligned}$$



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Since

$$\begin{aligned}b_1 - a_1 &= \frac{1}{2}(b_0 - a_0) \\b_2 - a_2 &= \frac{1}{2}(b_1 - a_1) = \frac{1}{4}(b_0 - a_0) \\&\vdots \\b_n - a_n &= \frac{1}{2^n}(b_0 - a_0)\end{aligned}$$

hence

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

Therefore

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

Since f is a continuous function, we have that

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



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On the other hand,

$$\begin{aligned} f(a_n)f(b_n) &\leq 0 \\ \Rightarrow \lim_{n \rightarrow \infty} f(a_n)f(b_n) &= f^2(z) \leq 0 \\ \Rightarrow f(z) &= 0 \end{aligned}$$

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

$$\begin{aligned} |z - c_n| &= \left| \lim_{n \rightarrow \infty} a_n - \frac{1}{2}(a_n + b_n) \right| \\ &= \left| \frac{1}{2} \left[\lim_{n \rightarrow \infty} a_n - b_n \right] + \frac{1}{2} \left[\lim_{n \rightarrow \infty} a_n - a_n \right] \right| \\ &\leq \max \left\{ \left| \lim_{n \rightarrow \infty} a_n - b_n \right|, \left| \lim_{n \rightarrow \infty} a_n - a_n \right| \right\} \\ &\leq |b_n - a_n| = \frac{1}{2^n} |b_0 - a_0|. \end{aligned}$$

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

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

then

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

Remark

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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

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

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

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That is, we solve

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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 $f(x^*) = x^* - g(x^*) = x^* - x^* = 0$

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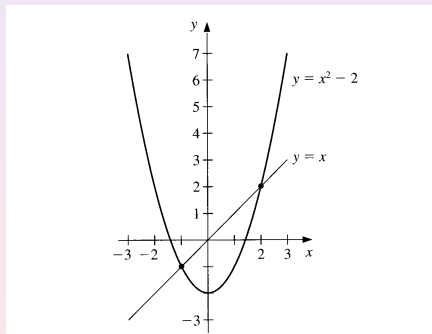
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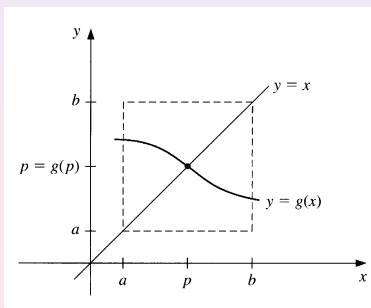
The function $g(x) = x^2 - 2$, for $-2 \leq x \leq 3$, has fixed points at $x = -1$ and $x = 2$ since

$$g(-1) = (-1)^2 - 2 = -1 \quad \text{and} \quad g(2) = 2^2 - 2 = 2.$$



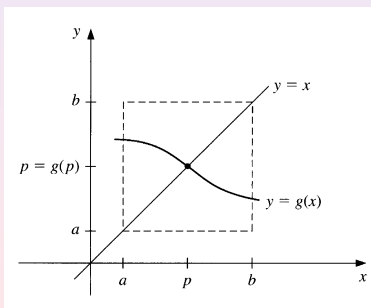
Theorem (Existence and uniqueness)

- 1 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**.



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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

$$h(a) = g(a) - a > 0 \quad \text{and} \quad h(b) = g(b) - b < 0.$$

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

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

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



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Uniqueness:

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

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

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Example

Show that the following function has a unique fixed point.

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

Solution: The Extreme Value Theorem implies that

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

That is $g(x) \in [-1/3, 0] \subset [-1, 1], \forall x \in [-1, 1]$.

Moreover, g is continuous and

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

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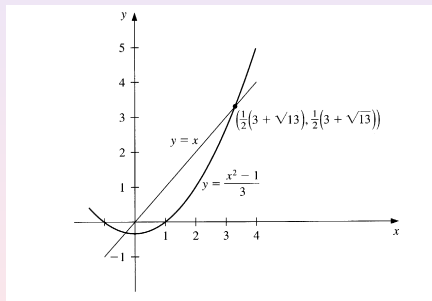
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By above theorem, g has a unique fixed point in $[-1, 1]$.



Let p be such unique fixed point of g . Then

$$\begin{aligned} p = g(p) = \frac{p^2 - 1}{3} &\Rightarrow p^2 - 3p - 1 = 0 \\ &\Rightarrow p = \frac{1}{2}(3 + \sqrt{13}). \end{aligned}$$



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,

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

then, since g is continuous,

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

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



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Fixed-point iteration

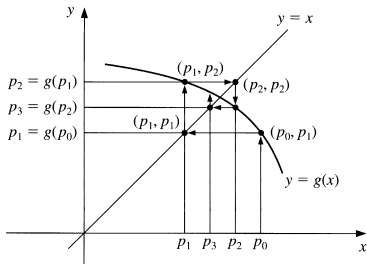
Given x_0 , tolerance TOL , maximum number of iteration M .

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

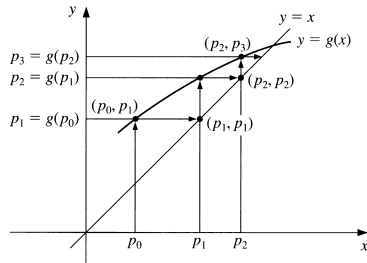
While $i \leq M$ and $|x - x_0| \geq TOL$

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

End While



(a)



(b)



Example

The equation

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

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

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

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

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



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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	1.03×10^8		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 $g \in [a, b]$ be such that $g(x) \in [a, b]$ for all $x \in [a, b]$. Suppose that g' exists on (a, b) and that $\exists k$ with $0 < k < 1$ such that

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

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

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

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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

$$|x_n - x| \leq k |x_{n-1} - x| \leq k^2 |x_{n-2} - x| \leq \cdots \leq k^n |x_0 - x|. \quad (1)$$

Since $0 < k < 1$, we have

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

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If g satisfies the hypotheses of above theorem, then

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

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$$|x_n - x| \leq \frac{k^n}{1 - k} |x_1 - x_0|, \quad \forall n \geq 1.$$

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

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It implies that

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Example

For previous example,

$$f(x) = x^3 + 4x^2 - 10 = 0.$$

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

$$g_1(1) = 6 \quad \text{and} \quad g_1(2) = -12,$$

so $g_1([1, 2]) \not\subseteq [1, 2]$. Moreover,

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

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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

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

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Suppose that $f : \mathbb{R} \rightarrow \mathbb{R}$ 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

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

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This sets the stage for the **Newton-Rapbson'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**

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



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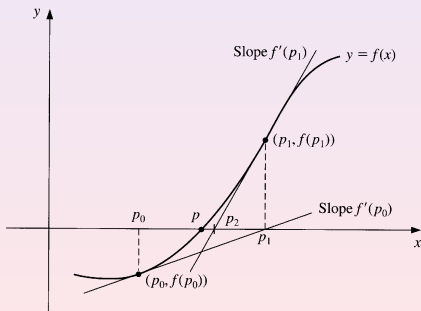
Given x_0 , tolerance TOL , maximum number of iteration M .

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

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

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

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

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

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

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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_n	$ 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

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

Proof: Define

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

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

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

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



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

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



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

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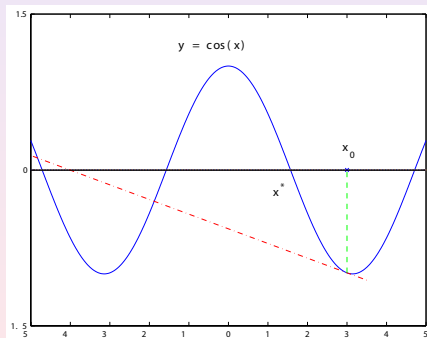
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Example

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 $x_1 = -4.01525$, $x_2 = -4.8526$, \dots , 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,

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

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

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

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

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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

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

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

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is then used as a new approximate x_n .



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Secant Method

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

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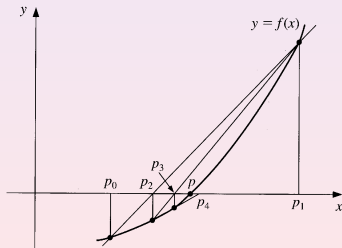
$$x = x_1 - y_1(x_1 - x_0)/(y_1 - y_0).$$

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

❶ Choose initial approximations x_0 and x_1 with $f(x_0)f(x_1) < 0$.

❷ $x_2 = x_1 - f(x_1)(x_1 - x_0)/(f(x_1) - f(x_0))$

❸ 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.,

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

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

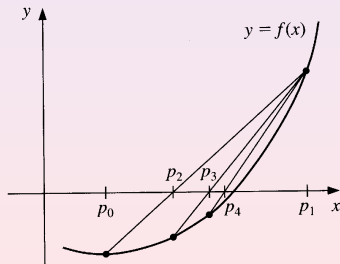
While $i \leq M$ and $|x - x_1| \geq TOL$

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

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

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Error analysis for iterative methods

Definition

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

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

then we say the **rate of convergence** is of **order α** .

We say that the rate of convergence is

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

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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

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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

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

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

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

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



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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

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

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

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

- Thus,

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

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



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- By the Fixed-Point Theorem, the sequence $\{x_n\}_{n=0}^{\infty}$ converges to x^* .
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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

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

Then $\exists \delta > 0$ such that

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

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For Newton's method,

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

It follows that $g'(x^*) = 0$. Hence Newton's method is locally quadratically convergent.



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

$$\begin{aligned}e_{k+1} &= x_{k+1} - x^* \\&= x_k - f(x_k) \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} - x^* \\&= \frac{1}{f(x_k) - f(x_{k-1})} [(x_{k-1} - x^*)f(x_k) - (x_k - x^*)f(x_{k-1})] \\&= \frac{1}{f(x_k) - f(x_{k-1})} (e_{k-1}f(x_k) - e_k f(x_{k-1})) \\&= e_k e_{k-1} \left(\frac{\frac{1}{e_k} f(x_k) - \frac{1}{e_{k-1}} f(x_{k-1})}{x_k - x_{k-1}} \cdot \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} \right)\end{aligned}$$



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

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

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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.,

$$\frac{|e_{k+1}|}{\eta |e_k|^\alpha} \rightarrow 1 \quad \text{as } k \rightarrow \infty.$$

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

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

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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



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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.

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- secant method: superlinear
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Each iteration of method requires

- secant method: one function evaluation
- Newton's method: two function evaluation, namely, $f(x_k)$ and $f'(x_k)$.

⇒ two steps of secant method are comparable to one step of Newton's method. Thus

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

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Remark

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



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Accelerating convergence

Aitken's Δ^2 method

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

For n sufficiently large,

$$\frac{y_{n+1} - y}{y_n - y} \approx \frac{y_{n+2} - y}{y_{n+1} - y}.$$

Then

$$(y_{n+1} - y)^2 \approx (y_{n+2} - y)(y_n - y),$$

so

$$y_{n+1}^2 - 2y_{n+1}y + y^2 \approx y_{n+2}y_n - (y_{n+2} + y_n)y + y^2$$



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$$\hat{y}_n = y_n - \frac{(y_{n+1} - y_n)^2}{(y_{n+2} - y_{n+1}) - (y_{n+1} - y_n)}. \quad (3)$$

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Example

The sequence $\{y_n = \cos(1/n)\}_{n=1}^{\infty}$ converges linearly to $y = 1$.

n	y_n	\hat{y}_n
1	0.54030	0.96178
2	0.87758	0.98213
3	0.94496	0.98979
4	0.96891	0.99342
5	0.98007	0.99541
6	0.98614	
7	0.98981	

- $\{\hat{y}_n\}_{n=1}^{\infty}$ converges more rapidly to $y = 1$ than $\{y_n\}_{n=1}^{\infty}$.



Definition

For a given sequence $\{y_n\}_{n=0}^{\infty}$, the forward difference Δy_n is defined by

$$\Delta y_n = y_{n+1} - y_n, \quad \text{for } n \geq 0.$$

Higher powers of Δ are defined recursively by

$$\Delta^k y_n = \Delta(\Delta^{k-1} y_n), \quad \text{for } k \geq 2.$$

The definition implies that

$$\Delta^2 y_n = \Delta(y_{n+1} - y_n) = \Delta y_{n+1} - \Delta y_n = (y_{n+2} - y_{n+1}) - (y_{n+1} - y_n).$$

So the formula for \hat{y}_n in (??) can be written as

$$\hat{y}_n = y_n - \frac{(\Delta y_n)^2}{\Delta^2 y_n}, \quad \text{for } n \geq 0.$$



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Theorem

Suppose $\{y_n\}_{n=0}^{\infty} \rightarrow y$ *linearly* and

$$\lim_{n \rightarrow \infty} \frac{y_{n+1} - y}{y_n - y} < 1.$$

Then $\{\hat{y}_n\}_{n=0}^{\infty} \rightarrow y$ *faster than* $\{y_n\}_{n=0}^{\infty}$ in the sense that

$$\lim_{n \rightarrow \infty} \frac{\hat{y}_n - y}{y_n - y} = 0.$$

- Aitken's Δ^2 method constructs the terms in order:

$$y_0, \quad y_1 = g(y_0), \quad y_2 = g(y_1), \quad \hat{y}_0 = \{\Delta^2\}(y_0), \quad y_3 = g(y_2), \\ \hat{y}_1 = \{\Delta^2\}(y_1), \quad \dots$$

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Steffensen's method (To find a solution of $y = g(y)$)

Given y_0 , tolerance TOL , maximum number of iteration M .

Set $i = 1$.

While $i \leq M$

Set $y_1 = g(y_0)$; $y_2 = g(y_1)$; $y = y_0 - (y_1 - y_0)^2 / (y_2 - 2y_1 + y_0)$.

If $|y - y_0| < Tol$, then STOP.

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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 $x_0 \in [x^* - \delta, x^* + \delta]$

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Zeros of polynomials and Müller's method

- Horner's method:

Let

$$\begin{aligned}P(x) &= a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1} + a_nx^n \\&= a_0 + x(a_1 + x(a_2 + \cdots + x(a_{n-1} + a_nx) \cdots)).\end{aligned}$$

If

$$\begin{aligned}b_n &= a_n, \\b_k &= a_k + b_{k+1}x_0, \text{ for } k = n-1, n-2, \dots, 1, 0,\end{aligned}$$

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$$\begin{aligned}P(x) &= a_0 + a_1x + a_2x^2 + \cdots + a_{n-1}x^{n-1} + a_nx^n \\&= a_0 + x(a_1 + x(a_2 + \cdots + x(a_{n-1} + a_nx) \cdots)).\end{aligned}$$

If

$$\begin{aligned}b_n &= a_n, \\b_k &= a_k + b_{k+1}x_0, \quad \text{for } k = n-1, n-2, \dots, 1, 0,\end{aligned}$$

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$$b_0 = a_0 + b_1x_0 = a_0 + (a_1 + b_2x_0)x_0 = \cdots = P(x_0).$$

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$$Q(x) = b_1 + b_2x + \cdots + b_nx^{n-1}.$$



Zeros of polynomials and Müller's method

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Differentiating $P(x)$ with respect to x gives

$$P'(x) = Q(x) + (x - x_0)Q'(x) \quad \text{and} \quad P'(x_0) = Q(x_0).$$

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$$x_{k+1} = x_k - \frac{P(x_k)}{Q(x_k)}, \quad \forall k = 0, 1, 2, \dots$$

Similarly, let

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Horner's method (Evaluate $y = P(x_0)$ and $z = P'(x_0)$)

Set $y = a_n; z = a_n$.

For $j = n - 1, n - 2, \dots, 1$

Set $y = a_j + yx_0; z = y + zx_0$.

End for

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If x_N is an approximate zero of P , then

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

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.



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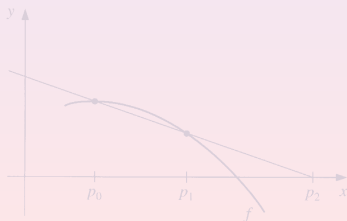


- Müller's method for complex root:

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 $(p_2, f(p_2))$.

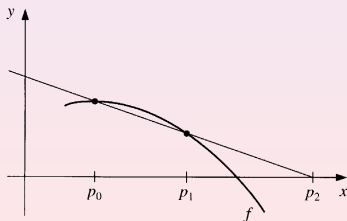


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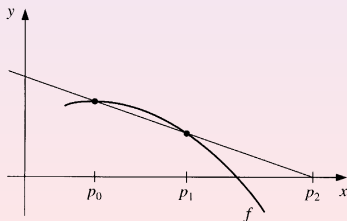


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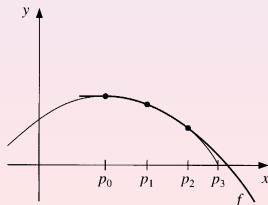
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Let

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

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

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

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

Choose

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

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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 $i \leq M$

$$\text{Set } b = \delta_2 + h_2 d; D = \sqrt{b^2 - 4f(p_2)d}.$$

If $|b - D| < |b + D|$, then set $E = b + D$ else set $E = b - D$.

$$\text{Set } h = -2f(p_2)/E; p = p_2 + h.$$

If $|h| < TOL$, then STOP.

$$\text{Set } p_0 = p_1; p_1 = p_2; p_2 = p; h_1 = p_1 - p_0; h_2 = p_2 - p_1;$$

$$\delta_1 = (f(p_1) - f(p_0))/h_1; \delta_2 = (f(p_2) - f(p_1))/h_2;$$

$$d = (\delta_2 - \delta_1)/(h_2 + h_1); i = i + 1.$$

End while

