

Nonlinear equations

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slides based on lecture notes/slides from L. Dede

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Plan

Numerical solution of a non-linear equation

Bisection method

- Convergence of the bisection method
- Stopping criterion and algorithm

Newton method

- Convergence
- Modified Newton method
- Stopping criterion and algorithm
- Inexact and quasi-Newton methods

Fixed point iterations

- Global convergence in an interval
- Local convergence in a neighborhood of α
- Stopping criterion for fixed point iterations
- Newton method as a fixed point iterations method
- Examples and their numerical solution
- Modified Newton method

Plan

Numerical solution of a non-linear equation

Bisection method

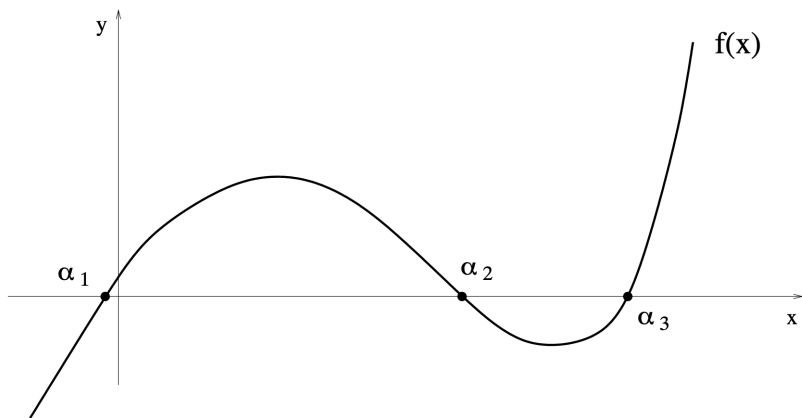
Newton method

Fixed point iterations

Nonlinear Equations

Objective: Approximate numerically the root of scalar (or vector) non-linear function $f(x)$, i.e.

find $\alpha \in \mathbb{R}$ such that $f(\alpha) = 0$ in the interval $I = (a, b) \subseteq \mathbb{R}$



Examples of nonlinear equations

Example 1 (Interest rates). We want to compute the mean interest rate I of a portfolio over several years. We invest $v = 1000$ CHF every year. After 5 years, we end up with $M = 6000$ CHF. The relation between M , v , I , and the number of years n is:

$$M = v \sum_{k=1}^n (1 + I)^k = v \frac{1 + I}{I} [(1 + I)^n - 1].$$

This can be rewritten as: find I such that

$$f(I) = M - v \frac{1 + I}{I} [(1 + I)^n - 1] = 0. \quad (1)$$

Therefore, we have to solve a nonlinear equation in I , for which we can't find an analytical solution.

Examples of nonlinear equations

Example 2 (State equation of a gas). We want to determine the volume V occupied by a gas at temperature T and pressure p . The state equation (i.e., the equation that relates p , V , and T) is:

$$\left(p + \frac{aN^2}{V^2}\right)(V - Nb) = kNT,$$

where a and b are two coefficients that depend on the specific gas, N is the number of molecules contained in the volume V , and k is the Boltzmann constant. We need, therefore, to solve a nonlinear equation whose root is V .



Plan

Numerical solution of a non-linear equation

Bisection method

- Convergence of the bisection method

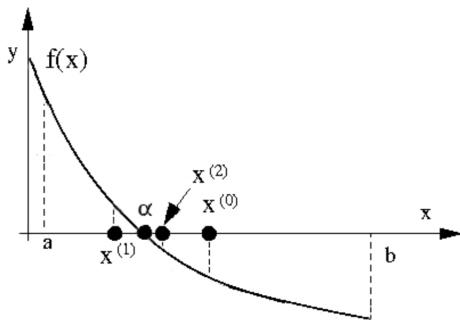
- Stopping criterion and algorithm

Newton method

Fixed point iterations

Bisection method (book chap 2.1)

- Compute the root of a **continuous function** f , i.e., the point α such that $f(\alpha) = 0$.
- Build a sequence $x^{(0)}, x^{(1)}, \dots, x^{(k)}$, with $x^{(0)}$ such that $\lim_{k \rightarrow \infty} x^{(k)} = \alpha$.



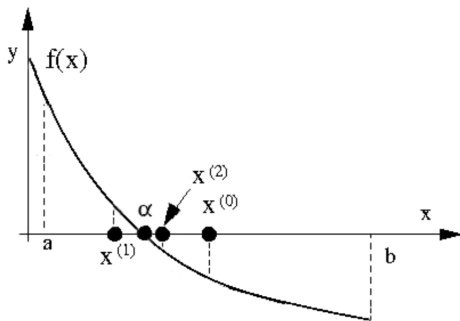
Foundation of the bisection method

Theorem 1 (Theorem 2.1 in book)

Zeros of a continuous function

Let $f(x)$ be a *continuous function* in $I = (a, b)$, that is $f \in C^0([a, b])$.

If $f(a)f(b) < 0$, then there exists at least one zero $\alpha \in I$ of the function $f(x)$.



Bisection method: algorithm

- Assume there exists an unique zero $\alpha \in (a, b)$ of $f \in C^0([a, b])$ and $f(a)f(b) < 0$
- Search α by recursively approximating it with the sequence of mid-points of subintervals $I^{(k)}$ of $I = (a, b)$ for which $f(x)$ changes sign

Bisection method: first step

Then start on $I^{(0)} = (a^{(0)}, b^{(0)}) = I = (a, b)$:

1. We set $a^{(0)} = a$, $b^{(0)} = b$, and $x^{(0)} = \frac{a^{(0)} + b^{(0)}}{2}$.

2. If $f(x^{(0)}) = 0$, then $x^{(0)}$ is the zero.

3. If $f(x^{(0)}) \neq 0$, then:

3.1 If $f(x^{(0)})f(a^{(0)}) > 0 \implies$ the zero $\alpha \in (x^{(0)}, b^{(0)})$, and we define:

$$a^{(1)} = x^{(0)}, \quad b^{(1)} = b^{(0)}, \quad x^{(1)} = \frac{a^{(1)} + b^{(1)}}{2}$$

3.2 If $f(x^{(0)})f(a^{(0)}) < 0 \implies$ the zero $\alpha \in (a^{(0)}, x^{(0)})$, and we define:

$$b^{(1)} = x^{(0)}, \quad a^{(1)} = a^{(0)}, \quad x^{(1)} = \frac{a^{(1)} + b^{(1)}}{2}$$

Continue recursively on $I^{(1)} = (a^{(1)}, b^{(1)})$, \dots $I^{(k)} = (a^{(k)}, b^{(k)})$ till convergence.

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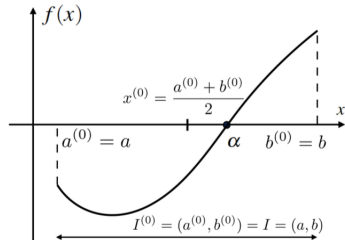
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Bisection method: example

Step 0.

$$I^{(0)} = (a^{(0)}, b^{(0)}) = I = (a, b) \text{ and}$$

$$x^{(0)} = \frac{a^{(0)} + b^{(0)}}{2} = \frac{a + b}{2}$$



Bisection method: example

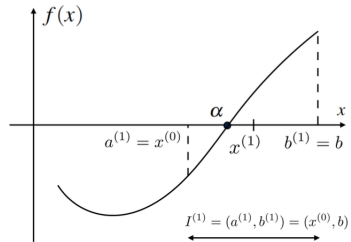
Step 1.

Since $f(x^{(0)}) f(b^{(0)}) < 0$:

$$a^{(1)} = x^{(0)}, b^{(1)} = b,$$

$I^{(1)} = (a^{(1)}, b^{(1)}) = (x^{(0)}, b)$, and

$$x^{(1)} = \frac{a^{(1)} + b^{(1)}}{2} = \frac{x^{(0)} + b}{2}.$$



Bisection method: example

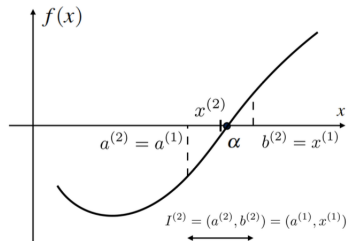
Step 2.

Since $f(x^{(1)}) f(a^{(1)}) < 0$:

$$a^{(2)} = a^{(1)}, b^{(2)} = x^{(1)},$$

$I^{(2)} = (a^{(2)}, b^{(2)}) = (a^{(1)}, x^{(1)})$, and

$$x^{(2)} = \frac{a^{(2)} + b^{(2)}}{2} = \frac{a^{(1)} + x^{(1)}}{2}.$$



Computational error of the bisection method

By repeating divisions of this type, we construct the sequence $x^{(0)}, x^{(1)}, \dots, x^{(k)}$ that satisfies for all k :

$$|I^{(k)}| := b^{(k)} - a^{(k)} \equiv \frac{|I^{(k-1)}|}{2} \quad \text{for all } k \geq 1,$$

$$|I^{(k)}| = \frac{|I^{(0)}|}{2^k} = \frac{b-a}{2^k} \quad \text{for all } k \geq 0.$$

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Bounding the computational error

Define computational error as

$$e^{(k)} = |x^{(k)} - \alpha|$$

and the error estimator as

$$\tilde{e}^{(k)} := |f^{(k+1)}| = \frac{b-a}{2^{k+1}}$$

Bounding the computational error

We have:

$$e^{(k)} \leq \tilde{e}^{(k)} := |I^{(k+1)}| = \frac{b-a}{2^{k+1}} \quad \text{for all } k \geq 0. \quad (2.1)$$

This implies that **the bisection method is convergent**; indeed

$$\lim_{k \rightarrow +\infty} e^{(k)} = 0$$

since $e^{(k)} \leq \tilde{e}^{(k)}$ for all $k \geq 0$ and

$$\lim_{k \rightarrow +\infty} \tilde{e}^{(k)} = \lim_{k \rightarrow +\infty} \frac{b-a}{2^{k+1}} = 0.$$

Convergence order

Definition 2 (Definition 2.1 in book)

An iterative method for the approximation of the zero α of the function $f(x)$ is convergent with order p if and only if

$$\lim_{k \rightarrow +\infty} \frac{|x^{(k+1)} - \alpha|}{|x^{(k)} - \alpha|^p} = \mu, \quad (1)$$

with $\mu > 0$ a real number independent of k , which is called the **asymptotic convergence factor**. In the case of linear convergence, i.e., for $p = 1$, we need $0 < \mu < 1$.

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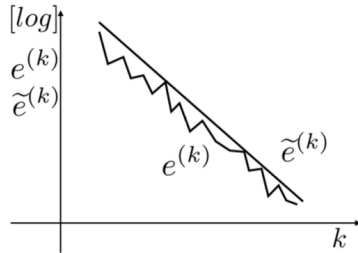
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No convergence order for bisection method

- The error may not be monotonically convergent, i.e., it is possible that $e^{(k+1)} > e^{(k)}$ for some $k \geq 0$;
- A convergence order cannot be established according to Eq. (1)

The sequence of error estimators $\{\tilde{e}^{(k)}\}$ is linearly convergent according to Eq. (1) with $\rho = 1$ and $\mu = \frac{1}{2}$; indeed:

$$\frac{\tilde{e}^{(k+1)}}{\tilde{e}^{(k)}} = \frac{(b-a)/2^{k+2}}{(b-a)/2^{k+1}} = \frac{1}{2} \quad \text{for all } k \geq 0.$$

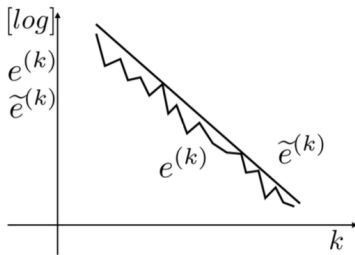


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Algorithm of bisection method

Algorithm 1 Bisection Method

Require: f , a , b , tol , k_{\max}

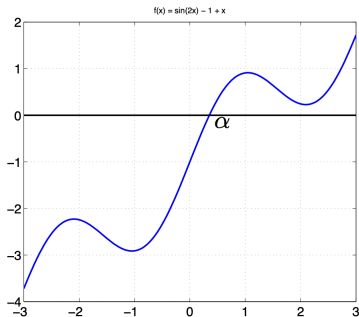
Assert: α

```
1: Set  $k = 0$ ,  $a^{(0)} = a$ ,  $b^{(0)} = b$ ,  $x^{(0)} = \frac{a+b}{2}$ ,  $ee^{(0)} = \frac{b-a}{2}$ 
2: while  $\tilde{e}^{(k)} > \text{tol}$  and  $k < k_{\max}$  do
3:   if  $f(x^{(k)}) = 0$  then
4:     Set  $\alpha = x^{(k)}$ 
5:     return
6:   else if  $f(a^{(k)})f(x^{(k)}) < 0$  then
7:     Set  $a^{(k+1)} = a^{(k)}$ ,  $b^{(k+1)} = x^{(k)}$ 
8:   else if  $f(b^{(k)})f(x^{(k)}) < 0$  then
9:     Set  $a^{(k+1)} = x^{(k)}$ ,  $b^{(k+1)} = b^{(k)}$ 
10:  end if
11:  Set  $x^{(k+1)} = \frac{a^{(k+1)} + b^{(k+1)}}{2}$ 
12:  Set  $\tilde{e}^{(k+1)} = \frac{b^{(k+1)} - a^{(k+1)}}{2}$ 
13:  Set  $k = k + 1$ 
14: end while
15: Set  $\alpha = x^{(k)}$ 
```

Bisection method: example

Example 3. We want to find the zero of the function $f(x) = \sin(2x) - 1 + x$.

We draw the graph of the function f using Matlab

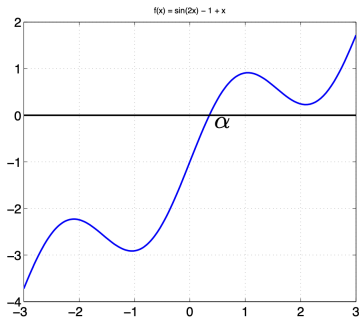


Bisection method: example

If we apply the bisection method in the interval $[-1, 1]$ with a tolerance of 10^{-8} and a maximum number of iterations $k_{\max} = 1000$, using the following command in Matlab:

```
[zero, res, niter] = bisection(f, -1, 1, 1e-8, 1000);
```

We find the value $\alpha = 0.352288462$ after 27 iterations.



Plan

Numerical solution of a non-linear equation

Bisection method

Newton method

- Convergence

- Modified Newton method

- Stopping criterion and algorithm

- Inexact and quasi-Newton methods

Fixed point iterations

Newton method (book, chap 2.2)

Let $f : \mathbb{R} \rightarrow \mathbb{R}$, $f \in C^0(I)$, be a differentiable function in $I = (a, b)$.

Let $x^{(0)}$ be an initial guess. Consider the equation of the tangent line to the curve $(x, f(x))$ at coordinate $x^{(k)}$ and with slope $f'(x^{(k)})$:

$$y(x) = f(x^{(k)}) + f'(x^{(k)})(x - x^{(k)}).$$

Newton method searches the new iterate $x^{(k+1)}$ at the intersection between the tangent line and the x-axis, i.e.,

$$y(x^{(k+1)}) = 0.$$

We deduce the Newton iterate:

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

provided $f'(x^{(k)}) \neq 0$ for all $k \geq 0$.

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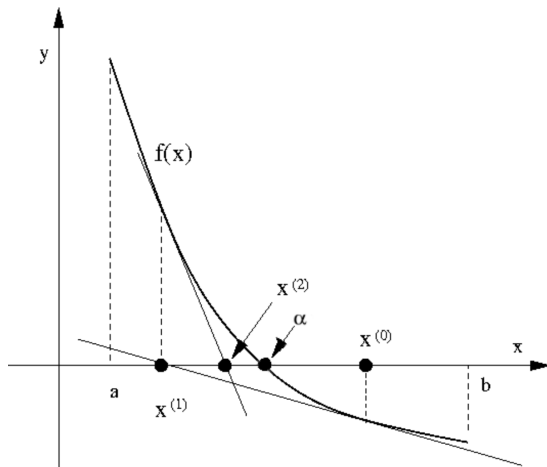
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Newton method

Starting from the point $x^{(0)}$, the root of f is obtained as the limit of the sequence $\{x^{(k)}\}_{k=0}^{\infty}$.



Newton method: algorithm

Algorithm 2 Newton Method

Require: f , $x^{(0)}$, tol , k_{\max}

Assert: α

- 1: Set $k = 0$
 - 2: **while** stopping criterion is false **do**
 - 3: Set $x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})}$
 - 4: Set $k = k + 1$
 - 5: **end while**
 - 6: Set $\alpha = x^{(k)}$
-

Newton method and Taylor expansion

Assume $f \in C^2(I)$, then the Taylor expansion of $f(x)$ around $x^{(k)}$ is:

$$f\left(x^{(k+1)}\right) = f\left(x^{(k)}\right) + f'\left(x^{(k)}\right)\left(x^{(k+1)} - x^{(k)}\right) + O\left(\left(x^{(k+1)} - x^{(k)}\right)^2\right)$$

If $f\left(x^{(k+1)}\right) = 0$, then the Newton method is the first-order approximation of the Taylor expansion of $f(x)$ around $x^{(k)}$;
in order to satisfy this assumption, one needs $x^{(k+1)} - x^{(k)}$ to be "small."

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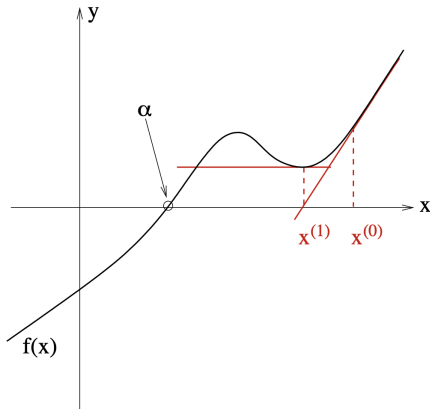
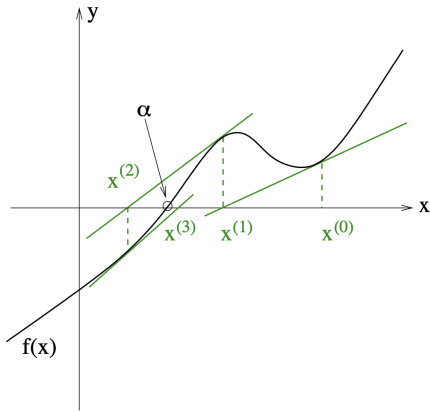
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Convergence

Does this method always converge?

- it depends on the **property of the function**;
- and on the **initial guess $x^{(0)}$** (should be "sufficiently" close to the zero α).



Convergence of the Newton method

Proposition 3 (Proposition 2.2 in book)

If

- $f \in C^1(I)$,
- $x^{(0)}$ is "sufficiently" close to $\alpha \in I$, and
- $f'(\alpha) \neq 0$,

then the Newton method is convergent to α , provided that $f'(x^{(k)}) \neq 0$ for all $k \geq 0$.

Convergence order of the Newton method

Proposition 4 (Proposition 2.3 in book)

Let I_α be a neighborhood of α . If $f \in C^2(I_\alpha)$, $x^{(0)}$ is "sufficiently" close to α , and $f'(\alpha) \neq 0$, then the Newton method is convergent with order 2 (quadratically) to α , provided that $f'(x^{(k)}) \neq 0$ for all $k \geq 0$. We have:

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{(x^{(k)} - \alpha)^2} = \frac{1}{2} \frac{f''(\alpha)}{f'(\alpha)},$$

following Eq. (1), $p = 2$ is the convergence order and $\mu = \frac{1}{2} \frac{f''(\alpha)}{f'(\alpha)}$ is the asymptotic convergence factor.

Proof.

Interpretation as a fixed point iterations method, next lecture !



Zero multiplicity

Definition 5 (Definition 2.2 in book)

Let $f \in C^m(I_\alpha)$, with $m \in \mathbb{N}$ such that $m \geq 1$.

The zero $\alpha \in I_\alpha$ is said to be of multiplicity m if $f^{(i)}(\alpha) = 0$ for all $i = 0, \dots, m-1$ and $f^{(m)}(\alpha) \neq 0$.

If $m = 1$, the zero α is called simple; otherwise, it is called multiple.

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Convergence order of Newton, zero multiple

Proposition 6 (Proposition 2.4 in book)

If $f \in C^2(I_\alpha) \cap C^m(I_\alpha)$ and $x^{(0)}$ is "sufficiently" close to the zero α of multiplicity $m > 1$, then the Newton method is convergent with order 1 (linearly) to α , provided that $f'(x^{(k)}) \neq 0$ for all $k \geq 0$. In particular, following Eq. (1), we have:

$$\lim_{k \rightarrow +\infty} \frac{|x^{(k+1)} - \alpha|}{|x^{(k)} - \alpha|} = \mu,$$

with $p = 1$ being the convergence order and $\mu \in (0, 1)$ the asymptotic convergence factor.

Modified Newton method

- Assume $f \in C^m(I_\alpha)$, with $\alpha \in I_\alpha$ and $m \geq 1$ the multiplicity of α .
- The modified Newton iterate is:

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

provided $f'(x^{(k)}) \neq 0$ for all $k \geq 0$.

- Requires knowledge or estimation of the multiplicity m

Convergence order of modified Newton method

Proposition 7 (Proposition 2.5 in book)

If $f \in C^2(I_\alpha) \cap C^m(I_\alpha)$, with $m \geq 1$ the multiplicity of the zero $\alpha \in I_\alpha$, and $x^{(0)}$ is “sufficiently” close to α ,

then modified Newton is convergent with order 2 to α , provided that $f'(x^{(k)}) \neq 0$ for all $k \geq 0$.

Stopping criterion for Newton

- Since α is unknown, the error $e^{(k)} = |x^{(k)} - \alpha|$ is also unknown
 \implies need to use estimators
- Criterion based on difference of successive iterates (discuss next lecture)
stop algorithm when $\tilde{e}^{(k)} < tol$, where

$$\tilde{e}^{(k)} = \begin{cases} |\delta^{(k-1)}| & \text{if } k > 0 \\ tol + 1 & \text{if } k = 0 \end{cases} \quad \text{with } \delta^{(k)} := x^{(k+1)} - x^{(k)} \quad \text{for } k \geq 0.$$

- Criterion based on residual

$$\tilde{e}^{(k)} = |r^{(k)}| \quad \text{with } r^{(k)} := f(x^{(k)}), \quad \text{for } k \geq 0$$

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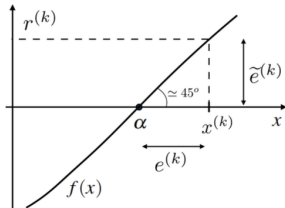
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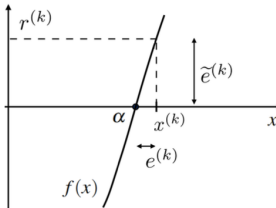
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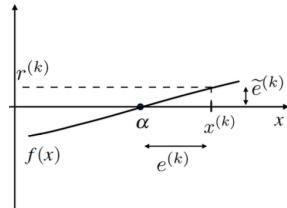
Satisfactory, $\tilde{e}^{(k)} \simeq e^{(k)}$



Unsatisfactory, $\tilde{e}^{(k)} \gg e^{(k)}$
(error overestimated)



Unsatisfactory, $\tilde{e}^{(k)} \ll e^{(k)}$
(error underestimated)



Newton method: algorithm

Algorithm 3 Newton's Method

Require: f , $x^{(0)}$, tol , k_{\max}

Assert: α

- 1: Set $k = 0$, $\tilde{e}^{(0)} = \text{tol} + 1$
 - 2: **while** $\tilde{e}^{(k)} > \text{tol}$ **and** $k < k_{\max}$ **do**
 - 3: Set $x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})}$
 - 4: Set $\tilde{e}^{(k+1)} = |x^{(k+1)} - x^{(k)}|$
 - 5: Set $k = k + 1$
 - 6: **end while**
 - 7: Set $\alpha = x^{(k)}$
-

Quasi-Newton methods

- Approximate $f'(x^{(k)})$ by computationally feasible $q^{(k)} \approx f'(x^{(k)})$.
- Quasi-Newton iterate is

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{q^{(k)}}, \quad k = 0, 1, 2, \dots$$

- Rope method:

$$q^{(k)} = \frac{f(b) - f(a)}{b - a}, \text{ for all } k \geq 0, \alpha \in (a, b)$$

- Secant method (order of convergence $p = 1.6$ if zero α is simple):

$$q^{(k)} = \frac{f(x^{(k)}) - f(x^{(k-1)})}{x^{(k)} - x^{(k-1)}}, \text{ for all } k \geq 1$$

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Rope and secant method at iterate $x^{(k)}$

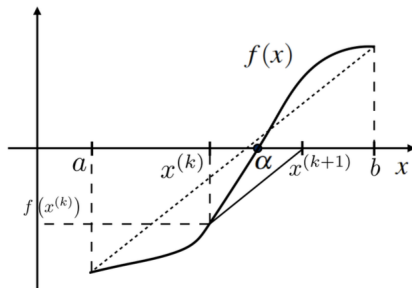
$$q^{(k)} = \frac{f(b) - f(a)}{b - a},$$

for all $k \geq 0, \alpha \in (a, b)$

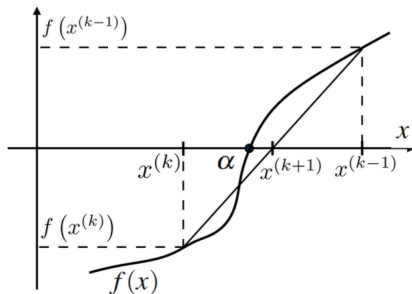
$$q^{(k)} = \frac{f(x^{(k)}) - f(x^{(k-1)})}{x^{(k)} - x^{(k-1)}},$$

for all $k \geq 1$

Rope method



Secant method



Plan

Numerical solution of a non-linear equation

Bisection method

Newton method

Fixed point iterations

- Global convergence in an interval

- Local convergence in a neighborhood of α

- Stopping criterion for fixed point iterations

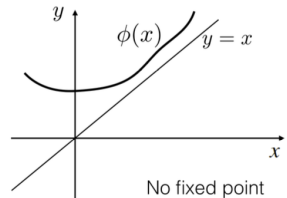
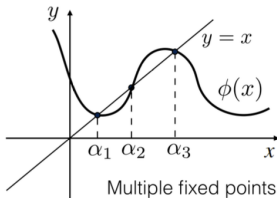
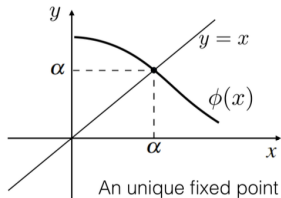
- Newton method as a fixed point iterations method

- Examples and their numerical solution

- Modified Newton method

Fixed point of a function

Definition 2.4: Given the iteration function $\phi : [a, b] \subseteq \mathbb{R} \rightarrow \mathbb{R}$, we say that $\alpha \in \mathbb{R}$ is a fixed point of ϕ if and only if $\phi(\alpha) = \alpha$.



Fixed point iterations (chap 2.3 in book)

Find the roots of nonlinear $f(x) = 0$ by solving the equivalent problem:

$$\phi : [a, b] \rightarrow \mathbb{R}. \quad x - \phi(x) = 0,$$

where ϕ must have the following property:

$$\phi(\alpha) = \alpha \text{ if and only if } f(\alpha) = 0$$

\implies Search for zeros of f by determining the fixed points of ϕ

Idea: It could be computed by the following algorithm:

$$x^{(k+1)} = \phi(x^{(k)}), \quad k \geq 0.$$

Indeed, if $x^{(k)} \rightarrow \alpha$ and if ϕ is continuous on $[a, b]$, then the limit α satisfies $\phi(\alpha) = \alpha$.

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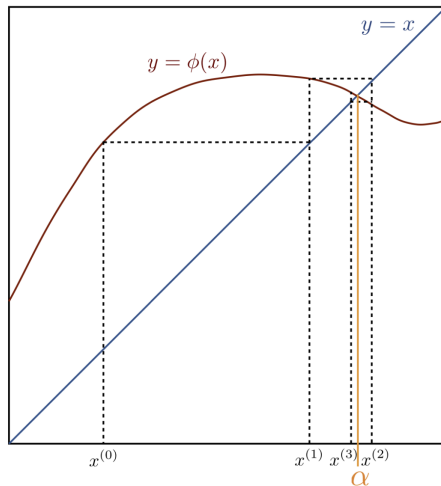
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Indeed, if $x^{(k)} \rightarrow \alpha$ and if ϕ is continuous on $[a, b]$, then the limit α satisfies $\phi(\alpha) = \alpha$.

Fixed point iterations (example)

Starting from the point $x^{(0)}$, the sequence $\{x^{(k)}\}$ converges to the fixed point α .



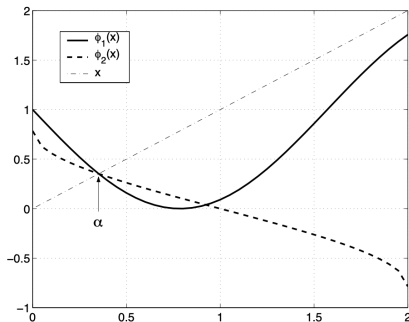
Example

Consider $f(x) = \sin(2x) - 1 + x = 0$. We can rewrite it in two different fashions. Using $f(\alpha) + \alpha = \alpha$,

$$x = \phi_1(x) = 1 - \sin(2x),$$

or setting $f(x) = 0$,

$$x = \phi_2(x) = \frac{1}{2} \arcsin(1 - x), \quad 0 \leq x \leq 1.$$



Global convergence in an interval

Proposition 8 (Proposition 2.7)

Consider iteration function $\phi : \mathbb{R} \rightarrow \mathbb{R}$ and the fixed point iterations.

1. If $\phi \in C^0([a, b])$ and $\phi(x) \in [a, b]$ for all $x \in [a, b]$, then there exists at least one fixed point $\alpha \in [a, b]$ of $\phi(x)$.
2. If, in addition, $\exists L \in [0, 1)$ s.t.

$$|\phi(x_1) - \phi(x_2)| \leq L|x_1 - x_2|$$

for all $x_1, x_2 \in [a, b]$, then the fixed point α is unique in $[a, b]$ and the fixed point iterations algorithm converges, for all the initial guesses $x^{(0)} \in [a, b]$:

$$\lim_{k \rightarrow +\infty} x^{(k)} = \alpha$$

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Proof of Proposition 2.7

1. Existence Consider the function $g(x) = \phi(x) - x$

Since $\phi(x) \in [a, b]$ for all $x \in [a, b]$, we have $g(a) = \phi(a) - a \geq 0$ and $g(b) = \phi(b) - b \leq 0$, for which $g(a)g(b) \leq 0$.

By applying the theorem of zeros of a continuous function, there exists at least a zero α of $g(x)$ in $[a, b]$; i.e. ϕ has at least one fixed point in $[a, b]$.

2. Uniqueness Indeed should two different fixed points exist, α_1 and α_2 , then

$$|\alpha_1 - \alpha_2| = |\phi(\alpha_1) - \phi(\alpha_2)| \leq L|\alpha_1 - \alpha_2| < |\alpha_1 - \alpha_2|$$

which cannot be. There exists a unique fixed point $\alpha \in [a, b]$ of ϕ

Proof of Proposition 2.7 (contd)

Convergence

Let $x^{(0)} \in [a, b]$ and $x^{(k+1)} = \phi(x^{(k)})$. We have

$$0 \leq |x^{(k+1)} - \alpha| = |\phi(x^{(k)}) - \phi(\alpha)| \leq L|x^{(k)} - \alpha| \leq \dots \leq L^{k+1}|x^{(0)} - \alpha|,$$

i.e.

$$\frac{|x^{(k)} - \alpha|}{|x^{(0)} - \alpha|} \leq L^k.$$

Because $L < 1$, for $k \rightarrow \infty$, we obtain

$$\lim_{k \rightarrow \infty} |x^{(k)} - \alpha| \leq \lim_{k \rightarrow \infty} L^k = 0.$$

So, for all $x^{(0)} \in [a, b]$, the sequence $\{x^{(k)}\}$ defined by $x^{(k+1)} = \phi(x^{(k)})$, $k \geq 0$, converges to α when $k \rightarrow \infty$.

Global convergence in an interval

Proposition 9 (Proposition 2.8)

If $\phi \in C^1([a, b])$, $\phi(x) \in [a, b]$ for all $x \in [a, b]$, and

$$|\phi'(x)| < 1 \quad \text{for all } x \in [a, b],$$

then there exists a unique fixed point $\alpha \in [a, b]$, and the fixed point iterations method converges for all $x^{(0)} \in [a, b]$ with at least linear order (i.e., order 1), that is:

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{x^{(k)} - \alpha} = \phi'(\alpha),$$

with $\phi'(\alpha)$ being the asymptotic convergence factor.

Note: If $0 < |\phi'(\alpha)| < 1$, then for any constant C s.t. $|\phi'(\alpha)| < C < 1$, if k is large enough, we have:

$$|x^{(k+1)} - \alpha| \leq C|x^{(k)} - \alpha|.$$

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Local convergence in a neighborhood of α

Theorem 10 (Theorem 2.9)

Lagrange Mean Value Theorem. *If the function $g \in C^1([a, b])$, then there exists $\xi \in (a, b)$ such that*

$$g(a) - g(b) = g'(\xi)(a - b).$$

Local convergence in a neighborhood of α

Proposition 2.10 — Ostrowski, local convergence in a neighborhood of the fixed point. If $\phi \in C^1(I_\alpha)$, with I_α a neighborhood of the fixed point α of $\phi(x)$, and

$$|\phi'(\alpha)| < 1,$$

then, if the initial guess $x^{(0)}$ is “sufficiently” close to α , the fixed point iterations method converges with at least linear order (i.e., order 1), that is:

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - a}{x^{(k)} - a} = \phi'(\alpha),$$

with $\phi'(\alpha)$ being the asymptotic convergence factor.

We only show that the method is at least linearly convergent.

Lagrange Theorem 2.9: there exists $\xi^{(k)} \in (a, x^{(k)})$ s.t.

$$x^{(k+1)} - \alpha = \phi(x^{(k)}) - \phi(\alpha) = \phi'(\xi^{(k)}) (x^{(k)} - \alpha),$$

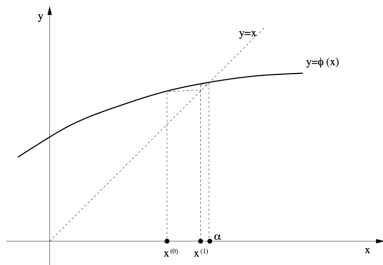
If $\lim_{k \rightarrow +\infty} x^{(k)} = \alpha$, then also $\lim_{k \rightarrow +\infty} \xi^{(k)} = \alpha$, and hence

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{x^{(k)} - \alpha} = \lim_{k \rightarrow +\infty} \phi'(\xi^{(k)}) = \phi'(\alpha).$$

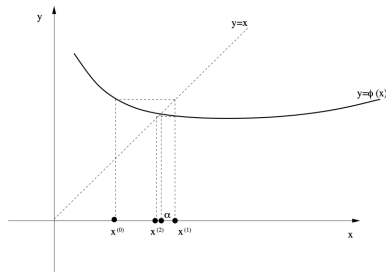
Examples

Some examples on how the value $\phi'(\alpha)$ influences the convergence.
Convergent cases:

$$0 < \phi'(\alpha) < 1,$$



$$-1 < \phi'(\alpha) < 0.$$

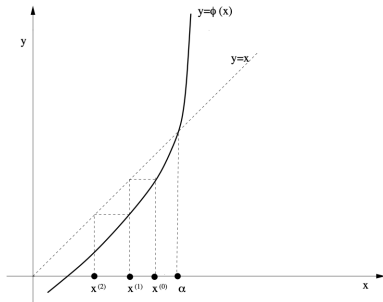


Examples

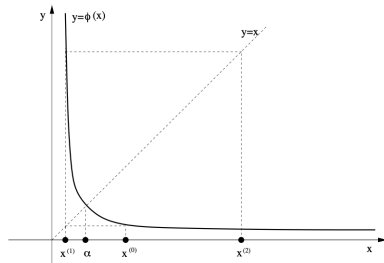
Some examples on how the value $\phi'(\alpha)$ influences the convergence.

Divergent cases:

$$\phi'(\alpha) > 1,$$



$$\phi'(\alpha) < -1.$$



- If $|\phi'(\alpha)| < 1$ and $x^{(0)}$ is “sufficiently” close to α ,
the method converges to α with at least linear order
- If $|\phi'(\alpha)| = 1$
the method may either converge or diverge
- If $|\phi'(\alpha)| > 1$ and $x^{(0)} \neq \alpha$,
the convergence of the method to α is impossible

Local convergence in a neighborhood of the fixed point

Proposition 11 (Proposition 2.11 in book)

If $\phi \in C^2(I_\alpha)$, with I_α a neighborhood of the fixed point α of $\phi(x)$, $\phi'(\alpha) = 0$, and $\phi''(\alpha) \neq 0$, then, if the initial guess $x^{(0)}$ is “sufficiently” close to α , the fixed point iterations method converges with **order 2**,

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{(x^{(k)} - \alpha)^2} = \frac{1}{2} \phi''(\alpha),$$

with $\frac{1}{2} \phi''(\alpha)$ being the asymptotic convergence factor.

Proof: Using the Taylor series for ϕ with $x = \alpha$, we have

$$x^{(k+1)} - \alpha = \phi(x^{(k)}) - \phi(\alpha) = \phi'(\alpha)(x^{(k)} - \alpha) + \frac{\phi''(\xi)}{2}(x^{(k)} - \alpha)^2,$$

where ξ is between $x^{(k)}$ and α . So, we have

$$\lim_{k \rightarrow \infty} \frac{x^{(k+1)} - \alpha}{(x^{(k)} - \alpha)^2} = \lim_{k \rightarrow \infty} \frac{\phi''(\xi)}{2} = \frac{\phi''(\alpha)}{2}.$$

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Generalization of the result

Proposition 12 (Proposition 2.12 in book)

If $\phi \in C^p(I_\alpha)$ for $p > 1$, with I_α a neighborhood of the fixed point α of $\phi(x)$, $\phi^{(i)}(\alpha) = 0$ for all $i = 1, \dots, p-1$, and $\phi^{(p)}(\alpha) \neq 0$, then, if the initial guess $x^{(0)}$ is “sufficiently” close to α , the fixed point iterations method converges with order p , that is:

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{(x^{(k)} - \alpha)^p} = \frac{1}{p!} \phi^{(p)}(\alpha),$$

with $\frac{1}{p!} \phi^{(p)}(\alpha)$ being the asymptotic convergence factor.

Example

Consider $f(x) = \sin(2x) - 1 + x = 0$. We have used the fixed point algorithms using the two functions ϕ_1 and ϕ_2 with initial value $x^{(0)} = 0.7$. Remember that both have the same fixed point α .

$$x = \phi_1(x) = 1 - \sin(2x),$$

$$x = \phi_2(x) = \frac{1}{2} \arcsin(1 - x), \quad 0 \leq x \leq 1.$$

```
>> [p1,res1,niter1] = fixpoint(phi1, 0.7, 1e-8, 1000);  
>> [p2,res2,niter2] = fixpoint(phi2, 0.7, 1e-8, 1000);
```

The fixed point algorithm with the first function does not converge, while with the second one it converges to $\alpha = 0.352288459558650$ in 44 iterations.

Indeed, $\phi_1'(\alpha) = -1.5237713$ and $\phi_2'(\alpha) = -0.65626645$.

Stopping criterion for fixed point iterations

- Estimate the error $e^{(k)} = |x^{(k)} - \alpha|$ by the difference of successive iterates
stop algorithm when $\tilde{e}^{(k)} < \text{tol}$, where

$$\tilde{e}^{(k)} = \begin{cases} |\delta^{(k-1)}| & \text{if } k \geq 1 \\ \text{tol} + 1 & \text{if } k = 0 \end{cases} \quad \text{with } \delta^{(k)} := x^{(k+1)} - x^{(k)} \quad \text{for } k \geq 0.$$

- If $\phi \in C^1(I_\alpha)$, by Lagrange Mean Value Theorem,

$$\exists \xi^{(k)} \text{ between } x^{(k)} \text{ and } \alpha \text{ s.t. } x^{(k+1)} - \alpha = \phi(x^{(k)}) - \phi(\alpha) = \phi'(\xi^{(k)})(x^{(k)} - \alpha)$$

- We also have:

$$x^{(k+1)} - \alpha = x^{(k+1)} - x^{(k)} + x^{(k)} - \alpha = \delta_k + x^{(k)} - \alpha$$

- We obtain

$$x^{(k)} - \alpha = -\frac{1}{1 - \phi'(\xi^{(k)})} \delta_k$$

Stopping criterion for fixed point iterations

$$e^{(k)} = x^{(k)} - \alpha = -\frac{1}{1 - \phi'(\xi^{(k)})} \delta_k$$

In a neighbourhood of α , $\phi'(\xi^{(k)}) \approx \phi'(\alpha)$

- if $\phi'(\alpha)$ is near to 1, the test is not satisfactory since $e^{(k)} \ll \tilde{e}^{(k+1)}$
- if $\phi'(\alpha) = 0$, the criterion is optimal (second order methods)
- if $\phi'(\alpha) \approx 0$, the criterion is satisfactory

Newton method as a fixed point iterations method

Consider the iterate of the Newton method

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

The Newton method is a fixed point method $x^{(k+1)} = \phi(x^{(k)})$ for the function:

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

We deduce properties of the Newton method from those of $\phi_N(x)$

Proposition 13 (Proposition 2.13 in book)

If $f \in C^m(I_\alpha)$, with I_α a neighborhood of the zero α and $m - 1$ the multiplicity of α , for the iteration function $x_N(x)$ of Eq. (2.8), we have

$$\phi'_N(\alpha) = 1 - \frac{1}{m}.$$

Proof. The proof is reported in Exercises Series 4

Convergence of Newton method, α simple zero

For Newton, the zero $\alpha \in I_\alpha$ is said to be of multiplicity m if $f^{(i)}(\alpha) = 0$ for all $i = 0, \dots, m-1$ and $f^{(m)}(\alpha) \neq 0$.

If $m = 1$, $f(\alpha) = 0$, and $f'(\alpha) \neq 0$, the zero α is called simple; otherwise, it is called multiple.

Remark 2.14 From Proposition 2.13, if α is a simple zero ($m = 1$), we have

$$\phi'_N(\alpha) = 1 - \frac{1}{m} = 1 - 1 = 0.$$

Convergence of Newton method, α simple zero

For Newton, the zero $\alpha \in I_\alpha$ is said to be of multiplicity m if $f^{(i)}(\alpha) = 0$ for all $i = 0, \dots, m-1$ and $f^{(m)}(\alpha) \neq 0$.

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$$\phi'_N(\alpha) = 1 - \frac{1}{m} = 1 - 1 = 0.$$

Convergence of Newton method, α simple zero

Corollary 14 (Corollary 2.14 from book)

If $f \in C^2(I_\alpha)$, α is a simple zero ($m = 1$), and $x^{(0)}$ is “sufficiently” close to α , then the Newton method converges with order 2 (quadratically). Indeed:

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{(x^{(k)} - \alpha)^2} = \frac{1}{2} \phi_N''(\alpha) = \frac{1}{2} \frac{f''(\alpha)}{f'(\alpha)},$$

where $f_N''(\alpha)$ is the second derivative of the Newton iteration function at α .

Proof.

Considering $\phi_N(x) = x - \frac{f(x)}{f'(x)}$, apply Propositions 2.11, 2.12, and Proposition 2.13. □

Convergence of Newton method, multiplicity $m > 1$

Corollary 15 (Corollary 2.15 from book)

If $f \in C^m(I_\alpha)$, α is a zero of multiplicity $m > 1$, and $x^{(0)}$ is “sufficiently” close to α , then the Newton method converges with order 1 (linearly),

$$\lim_{k \rightarrow +\infty} \frac{x^{(k+1)} - \alpha}{x^{(k)} - \alpha} = \phi'_N(\alpha) = 1 - \frac{1}{m} \neq 0.$$

Proof.

Considering $\phi_N(x) = x - \frac{f(x)}{f'(x)}$, apply Proposition 2.10 and Proposition 2.13. □

Examples of nonlinear equations

Example 1 (Interest rates). We want to compute the mean interest rate I of a portfolio over several years. We invest $v = 1000$ CHF every year. After 5 years, we end up with $M = 6000$ CHF. The relation between M , v , I , and the number of years n is:

$$M = v \sum_{k=1}^n (1 + I)^k = v \frac{1 + I}{I} [(1 + I)^n - 1].$$

This can be rewritten as: find I such that

$$f(I) = M - v \frac{1 + I}{I} [(1 + I)^n - 1] = 0. \quad (1)$$

Therefore, we have to solve a nonlinear equation in I , for which we can't find an analytical solution.

Example 1 (Interest rates)

We draw the graph of

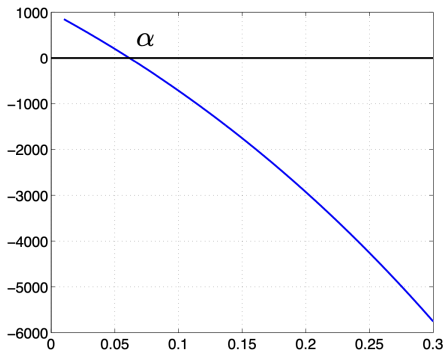
$$f(I) = M - v \frac{1 + I}{I} [(1 + I)^n - 1]$$

on the interval $[0.01; 0.3]$ with $M = 6000$, $v = 1000$, and $n = 5$:

```
>> f=@(x) 6000-1000*(1+x).*((1+x).^5 - 1)./x;
```

```
>> I = [0.01:0.001:0.3];
```

```
>> grid on;plot(I,feval(f,x));
```



Example 1 (Interest rates)

- The root of f is between 0.05 and 0.1.
- Apply the bisection method on the interval $[0.05, 0.1]$ with a tolerance 10^{-5}

```
>> [zero,res,niter]=bisection(f,0.05,0.1,1e-5,1000);
```

- The approximate solution after 12 iterations is $x = 0.061407470703125$.
- Apply the Newton method with initial guess $x^{(0)} = 0.05$

```
>> df=@(x) 1000*((1+x).^5.*(1-5*x) - 1)./(x.^2);  
>> [zero,res,niter]=newton(f,df,.05,1e-5,1000);
```

- The result is approximately the same, but we need only 3 iterations
- The interest rate is 6.14%.

Examples of nonlinear equations

Example 2 (State equation of a gas). We want to determine the volume V occupied by a gas at temperature T and pressure p . The state equation (i.e., the equation that relates p , V , and T) is:

$$\left(p + \frac{aN^2}{V^2}\right)(V - Nb) = kNT,$$

where a and b are two coefficients that depend on the specific gas, N is the number of molecules contained in the volume V , and k is the Boltzmann constant.



We consider the carbon dioxide (CO_2), for which $a = 0.401 \text{ Pa} \cdot \text{m}^6$ and $b = 42.7 \times 10^{-6} \text{ m}^3$.

We search the volume occupied by $N = 1000$ molecules of CO_2 at temperature $T = 300 \text{ K}$ and pressure $p = 3.5 \times 10^7 \text{ Pa}$.

We know that the Boltzmann constant is $k = 1.3806503 \times 10^{-23} \text{ Joule} \cdot \text{K}^{-1}$.

Example 2 (State equation of a gas)

We draw the graph of the function

$$f(V) = \left(p + \frac{aN^2}{V^2} \right) (V - Nb) - kNT$$

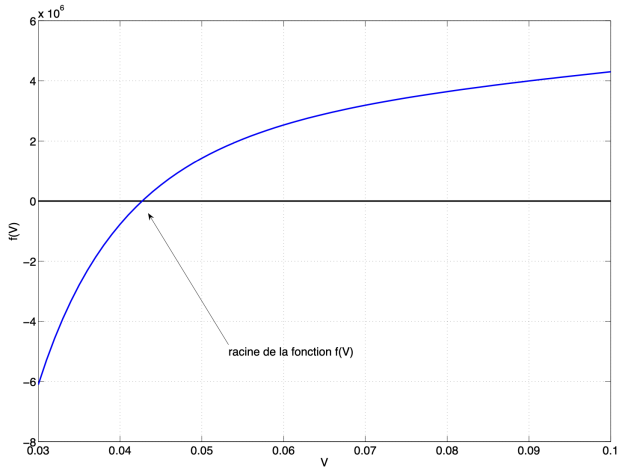
for $V > 0$. We do not consider $V < 0$ (it does not have physical sens), because V is the volume of gas.

We use the commands in Matlab:

```
>> a=0.401; b=42.7e-6; p=3.5e7; T=300; N=1000; k=1.3806503e-23;  
>> f =@(x,p,T,a,b,N,k) (p+a*((N./x).^2)).*(x-N*b)-k*N*T;  
>> x=[0.03:0.001:0.1];  
>> plot(x,f(x,p,T,a,b,N,k))  
>> grid on
```

Example 2 (State equation of a gas)

We obtain the graph of the function $f(V)$:



Example 2 (State equation of a gas)

We see that there is a zero for $0.03 < V < 0.1$. If we apply the bisection method on the interval $[0.03, 0.1]$ with a tolerance of 10^{-12} :

```
[zero, res, niter] = bisection(f, 0.03, 0.1, 1 × 10-12, 1000, p, T, a, b, N, k);
```

then we find, after 36 iterations, the value $V = 0.0427$.

If we use the Newton method with the same tolerance, starting from the initial point $x^{(0)} = 0.03$,

```
>> df = @(x,p,T,a,b,N,k) -2*a*N^2/(x^3)*(x - N*b) + (p + a*((N)/(x)).^2);  
>> [zero,res,niter] = newton(f, df, 0.03, 1e-12, 1000, p, T, a, b, N, k);  
then we find the same solution after 6 iterations.
```

The conclusion is that the volume V occupied by the gas is 0.0427 m^3 .

Modified Newton method

For the iterate of the modified Newton method

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

associate the iteration function $\phi_{mN}(x)$ defined as:

$$\phi_{mN}(x) = x - m \frac{f(x)}{f'(x)},$$

where m is the multiplicity of the zero α .

Convergence of modified Newton

Proposition 16 (Proposition 2.16 from book)

If $f \in C^m(I_\alpha)$, with I_α a neighborhood of the zero α and $m - 1$ the multiplicity of α , for the iteration function $\phi_{mN}(x)$, we have

$$\phi'_{mN}(\alpha) = 1 - \frac{m}{m} = 0 \text{ for all } m \geq 1.$$

Proof.

The result follows analogously to that of Proposition 2.13, reported in exercise series □

Corollary 17 (Corollary 2.17 from book)

If $f \in C^2(I_\alpha) \cap C^m(I_\alpha)$, α is a zero of multiplicity $m - 1$, and $x^{(0)}$ is “sufficiently” close to α , then the modified Newton method converges with order 2 (quadratically).

Proof.

From Proposition 2.11, 2.12, 2.16. □

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From Proposition 2.11, 2.12, 2.16. □