

## TMA4125 Matematikk 4N

Numerical Solution of Nonlinear Equations

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#### Introduction.

We know that quadratic equations of the form

$$ax^2 + bx + c = 0$$

have the roots (or solutions) 
$$r^{\pm} = rac{-b \pm \sqrt{d^2 - 4ac}}{2a}$$

More generally, for a given function f we want to consider the equation

$$f(x) = 0. \tag{1}$$

A solution  $x^*$  of (1) is called a root to the equation.

**Challenge.** In many applications, we encounter equations for which we do not have a simple solution formula as for quadratic functions. In fact, an analytical solution might not even exist!  $\Rightarrow$  Develop numerical techniques to solve (1).



#### Scalar and systems of equations

We consider scalar equations first, i.e.  $f : \mathbb{R} \to \mathbb{R}$ , or with just one equation and variable, for example

$$x^3 + x^2 - 3x = 3.$$

Later we will also consider systems of equations, for example

$$xe^{y} = 1,$$
$$-x^{2} + y = 1.$$

We can write this also as a functions  $f : \mathbb{R}^n \to \mathbb{R}^n$ , which in this example would be n = 2 and

$$f(x,y) = \begin{pmatrix} xe^y - 1 \\ -x^2 + y - 1 \end{pmatrix}$$
 and we want to solve  $f(x,y) = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ 

#### **Scalar equations**

Let  $f : [a, b] \to \mathbb{R}$  be a continuous function on some interval [a, b].

#### **Goal.** Find a zero or root of *f*.

Let's get an intuition by plotting the function (try yourself in Python). For example  $f(x) = x^3 + x^2 - 3x - 3$  from last slide looks like





### **Existence and uniqueness of solutions**

**Theorem.** Let  $f \in C[a, b]$  be given.

- **1.** if f(a) and f(b) are of different sign, then there exists at lesat one  $r \in (a, b)$  such that f(r) = 0Trick. the condition of different sign is f(a)f(b) < 0 in short.
- **2.** The solution is unique if  $f \in C^1[a, b]$  and f'(x) < 0 or f'(x) > 0 for all  $x \in (a, b)$ .



#### **Bisection method**

First part of the theorem already provides a very intuitive algorithm: Divide [a, b] into two parts [a, c], [c, b], check in which half the zero is and continue there.

#### Algorithm.

**Input** a function *f* and an interval, such that f(a)f(b) < 0

**1.** Set 
$$a^{(0)} = a$$
,  $b^{(0)} = b$ 

**2.** For 
$$k = 0, 1, 2, \dots$$

$$c^{(k)} = \frac{a^{(k)} + b^{(k)}}{2}$$

Set the next interval to  

$$[a^{(k+1)}, b^{(k+1)}] = \begin{cases} [a^{(k)}, c^{(k)}] & \text{if } f(a^{(k)})f(c^{(k)}) < 0 \\ [c^{(k)}, b^{(k)}] & \text{if } f(c^{(k)})f(b^{(k)}) < 0 \end{cases}$$

Idea.  $c^{(k)}$  approximates the root r, note that  $|c^{(k)} - r| \le \frac{b^{(k)} - a^{(k)}}{2}$ We can stop if  $\frac{b^{(k)} - a^{(k)}}{2}$  is small enough or if  $f(c^{(k)})$  is zero.



#### **Bisection method – exercises and summary**

#### Exercises.

- Choose appropriate intervals and find the other two roots of f
- ► Compute the solution(s) of x<sup>2</sup> + sin(x) 0.5 = 0 using the bisection method
- Given the interval [1.5, 2]. How many iterations are required to guarantee that the error  $|c^{(k)} r|$  is less than  $10^{-4}$

#### Summary.

The bisection method is very robust but not particularly fast.



#### **Fix point iterations**

A major class of iteration schemes are the so-called fix point iterations.

**General Idea.** Given an equation f(x) = 0 with root r. Rewrite the equation

f(x)=0

to a a fix point form x = g(x), i. e. construct a function g such that the root  $x^*$  of f is a fixed point of g, that is  $x^*$  satisfies

$$x^* = g(x^*).$$

#### Algorithm.

**Input** Given a function g and a starting value  $x^{(0)}$ 

**1.** For k = 0, 1, 2, ... compute

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

## Example. Fixed point equation

We rewrite

$$f(x) = x^3 + x^2 - 3x - 3 = 0$$
 to  $x = \frac{x^3 + x^2 + 3}{3} = g(x)$ .

Interpretation: Fixed points are intersections of g(x) with y = x.



## **Example. Fixed point equation (cont.)**

**Exercise.** Repeat the experiment with  $x^{(0)} = 1.5$  and

$$g_2(x) = rac{-x^2 + 3x + 3}{x^2}, \quad g_3(x) = \sqrt[3]{3 + 3x - x^2}, \quad g_4(x) = \sqrt{rac{3 + 3x - x^2}{x}}$$

Interpretation: Fixed points are intersections of g(x) with y = x.



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## Existence and uniqueness for fix point iteration

We can adapt the previous Theorem on the equation f(x) = x - g(x) = 0

**Theorem.** Let  $g \in C[a, b]$  **1.** if a < g(x) < b for all  $x \in (a, b)$  then g has at least one fixed point  $x^*$ **2.** if  $g \in C^1[a, b]$  and |g'(x)| < 1 for all  $x \in [a, b]$  then the fixed point  $x^*$  is unique.

We write the assumption a < g(x) < b for all  $x \in [a, b]$  as  $g([a, b]) \subset (a, b)$ .

#### Convergence of fix the point iteration

We denote the error after *k* iterations as  $e^{(k)} = x^* - x^{(k)}$ .

The iteration converges if  $e^{(k)} \rightarrow 0$  as  $k \rightarrow \infty$ . Under which conditions is this the case?

Here's a trick we will use: For any k we have

$$x^{(k+1)} = g(x^{(k)}),$$
 the iterations  
 $x^* = g(x^*)$  the fixed point.

Now use the last Theorem from the preliminaries (that generalised Rolle's theorem)

$$|e^{(k+1)}| = |x^* - x^{(k+1)}| = |g(x^*) - g(x^{(k)})| = |g'(\xi_k)| \cdot |x^* - x^{(k)}| = |g'(\xi_k)| \cdot |e_k|$$

where  $\xi_k$  is some unknown value between  $x^{(k)}$  (known) and  $x^*$  (unknown).

### The fix point theorem

**Theorem.** If there is an interval [a, b] such that  $g \in C^1[a, b]$ ,  $g([a, b]) \subset (a, b)$  and there exist a positive constant L < 1 such that  $|g'(x)| \le L < 1$  for all  $x \in [a, b]$ , then

- g has a unique fixed point r in (a, b).
- ► The fixed point iterations x<sup>(k+1)</sup> = g(x<sup>(k)</sup>) converges towards x<sup>\*</sup> for all starting values x<sup>(0)</sup> ∈ [a, b].

The property  $g([a, b]) \subset (a, b)$  guarantees that if  $x^{(0)} \in [a, b]$  then  $x^{(k)} \in (a, b)$ , for k = 1, 2, ...The condition  $|g'(x)| \le L < 1$  guarantees convergence towards the unique fixed point  $x^*$ , since

$$|e^{(k+1)}| \leq L|e^{(k)}| \quad \Rightarrow \quad |e^{(k)}| \leq L^k \, |e^{(0)}| o 0 \quad ext{as } k o \infty.$$

since  $L^k \to 0$  as  $k \to \infty$  due to L < 1 (also called linear convergence).

#### Example. The fix point interation (cont. II)



**Exercise.** How large can we make [-0.7, 1.3] still having convergence? Check the same conditions for  $g_2, g_3, g_4$ .

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#### Newton's method

Since a small  $g'(x^*)$  is preferable and we can choose  $g \Rightarrow$  can we choose g such that  $g'(x^*) = 0$ ?

We use  $x^{(k)} = x^* - e^{(k)}$  and a Taylor expansion (around  $x^*$ )

$$egin{aligned} e^{(k+1)} &= x^* - x^{(k+1)} = g(x^*) - g(x^{(k)}) = g(x^*) - g(x^* - e^{(k)}) \ &= -g'(x^*)e^{(k)} + rac{1}{2}g''(\xi_k) ig(e^{(k)}ig)^2 \end{aligned}$$

Let's choose  $g'(x^*) = 0$  and assume that we have a constant M such that  $|g''(x)|/2 \le M$ . Then

$$|e^{(k+1)}| \le M |e^{(k)}|^2$$

Also known as quadratic convergence.

#### How to find our favourite fix point equation.

**Question.** Given an equation

$$f(x)=0$$

with an unknown solution  $x^*$ . Can we find a g with fixed point  $x^*$  and  $g'(x^*) = 0$ ? **Idea.** Note that for any h(x) we have due to  $f(x^*) = 0$  that

$$g(x) = x - h(x)f(x)$$

has a fixed point  $x^*$ . The derivate reads

$$g(x) = 1 - h'(x)f(x) - h(x)f'(x)$$

and at the fixed point

$$g'(x^*) = 1 - h(x^*)f'(x^*).$$

Choose h(x) = 1/f'(x) we achieve  $g'(x^*) = 0$ .



#### Newton's method. Algorithm

#### Algorithm.

**Input** a function *f*, its derivative f', and a start point  $x^{(0)}$ .

**1.** For 
$$k = 0, 1, 2, ...$$
 compute

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})}{f'(x^{(k)})}$$

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#### **Error analysis**

We constructed the method to give quadratic convergence

 $|e^{(k+1)}| \leq M |e^{(k)}|^2,$ 

where  $e^{(k)} = x^* - x^{(k)}$ .

But under which conditions can we say something about the size of M?

Let's compare the Taylor series around  $x^*$  with Newton's method

$$0 = f(x^{(k)}) + f'(x^{(k)})(x^* - x^{(k)}) + \frac{1}{2}f''(\xi_k)(x^* - x^{(k)})^2 \quad \text{(Taylor series)}$$
  
$$0 = f(x^{(k)}) + f'(x^{(k)})(x^{(k+1)} - x^{(k)}), \qquad \text{(Newton's method)}$$

where  $\xi_k$  is between  $x^*$  and  $x^{(k)}$ . Subtracting both yields

$$f'(x^{(k)})(x^*-x^{(k+1)})+rac{1}{2}f''(\xi_k)(x^*-x^{(k)})^2=0 \quad \Rightarrow \quad e^{(k+1)}=-rac{1}{2}rac{f''(\xi_k)}{f'(x^{(k)})}(e^{(k)})^2$$

 $\Rightarrow$  we need  $f'(x^{(k)}) \neq 0$ ,  $f \in C^2[a, b]$  and  $x^{(0)}$  sufficiently close to  $x^*$ .

### **Convergence of Newton's method**

**Theorem.** Assume that the function *f* has a root  $x^*$ , and let  $I_{\delta} = [x^* - \delta, x^* + \delta]$  for some  $\delta > 0$ . Assume further that

There is a 
$$M > 0$$
 such that  $\left| \frac{f''(y)}{f'(x)} \right| \le 2M$ , for all  $x, y \in I_{\delta}$ .

In this case, Newton's method converges quadratically,

 $|e^{(k+1)}| \le M |e^{(k)}|^2$ 

for all starting values satisfying  $|x^{(0)} - x^*| \le \min\{1/M, \delta\}$ .

#### Exercises.

- Repeat the example with different  $x^{(0)}$  to find the two other roots.
- Verify quadratic convergence numerically (remember the EOC!).
- Solve x(1 cos(x)) = 0, both by the bisection and Newton's method, x<sup>(0)</sup> = 1. What difference do you observe?



#### Systems of equations

Instead of one function *f* with one variable *x* we now consider

#### Systems of nonlinear equations.

$$f_1(x_1, x_2, ..., x_n) = 0$$
  

$$f_2(x_1, x_2, ..., x_n) = 0$$
  
:  

$$f_n(x_1, x_2, ..., x_n) = 0$$

We can write this in short as

$$\boldsymbol{f}(\boldsymbol{x})=0,$$

where  $\boldsymbol{f} : \mathbb{R}^n \to \mathbb{R}^n$ .

#### Example.

**Example.** Consider the two equations

$$x^{3} - y + \frac{1}{4} = 0$$
$$x^{2} + y^{2} - 1 = 0$$

#### Interpretation.

Rewrite first equation to  $y = x^3 + \frac{1}{4} \Rightarrow$  solutions lie on this graph. Second equation means  $1 = x^2 + y^2 \Rightarrow$  point with distance 1 from origin



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#### Towards Newton's method for systems of equations I

**Idea.** Extend fixed point iterations to systems of equations. We concentrate on Newton's method and the case n = 2:

f(x, y) = 0g(x, y) = 0

to avoid getting lost in indices.

**Notation.** We denote a solution (root) to these by  $\mathbf{r} = (x^*, y^*)^{\mathrm{T}}$ . Let  $\hat{\mathbf{x}} = (\hat{x}, \hat{y})^{\mathrm{T}}$  that approximates  $\mathbf{r}$ .

 $\Rightarrow$  Search for a better approximation by linearizing f(x) = 0. In other words: use the multivariate Taylor expansion around  $\hat{x}$ 

$$f(x,y) = f(\hat{x},\hat{y}) + \frac{\partial f}{\partial x}(\hat{x},\hat{y})(x-\hat{x}) + \frac{\partial f}{\partial y}(\hat{x},\hat{y})(y-\hat{y}) + \dots$$
$$g(x,y) = g(\hat{x},\hat{y}) + \frac{\partial g}{\partial x}(\hat{x},\hat{y})(x-\hat{x}) + \frac{\partial g}{\partial y}(\hat{x},\hat{y})(y-\hat{y}) + \dots$$

where we omit higher order terms (in "..."), which are small for  $\pmb{x} \approx \hat{\pmb{x}}$ .

#### Towards Newton's method for systems of equations II

Idea. Ignore the higher order terms and solve

$$f(\hat{x}, \hat{y}) + \frac{\partial f}{\partial x}(\hat{x}, \hat{y})(x - \hat{x}) + \frac{\partial f}{\partial y}(\hat{x}, \hat{y})(y - \hat{y}) = 0$$
$$g(\hat{x}, \hat{y}) + \frac{\partial g}{\partial x}(\hat{x}, \hat{y})(x - \hat{x}) + \frac{\partial g}{\partial y}(\hat{x}, \hat{y})(y - \hat{y}) = 0$$

for *x* and *y* as a better approximation (or precise: next iterate). More compact we can write

$$f(\hat{\boldsymbol{x}}) + J(\hat{\boldsymbol{x}})(\boldsymbol{x} - \hat{\boldsymbol{x}}) = 0,$$

where J(x) denotes the Jacobian of f given by

$$J(\mathbf{x}) = \begin{pmatrix} \frac{\partial f}{\partial x}(x, y) & \frac{\partial f}{\partial y}(x, y) \\ \frac{\partial g}{\partial x}(x, y) & \frac{\partial g}{\partial y}(x, y) \end{pmatrix}$$



## Newton's method for systems of equations

#### Algorithm.

**Input** A function f(x), its Jacobian J(x) and a starting value  $x^{(0)}$ .

- **1.** For k = 0, 1, 2, ...
  - **1.1** Solve the linear system  $J(\mathbf{x}^{(k)})\Delta^{(k)} = -\mathbf{f}(\mathbf{x}^{(k)})$ **1.2** Set  $\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \Delta^{(k)}$

**Note.** This can be generalized to *n* equations with *n* unknowns, where the Jacobian reads

$$J(\mathbf{x}) = \begin{pmatrix} \frac{\partial f_1}{\partial x_1}(\mathbf{x}) & \frac{\partial f_1}{\partial x_2}(\mathbf{x}) & \cdots & \frac{\partial f_1}{\partial x_n}(\mathbf{x}) \\ \frac{\partial f_2}{\partial x_1}(\mathbf{x}) & \frac{\partial f_2}{\partial x_2}(\mathbf{x}) & \cdots & \frac{\partial f_2}{\partial x_n}(\mathbf{x}) \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1}(\mathbf{x}) & \frac{\partial f_n}{\partial x_2}(\mathbf{x}) & \cdots & \frac{\partial f_n}{\partial x_n}(\mathbf{x}) \end{pmatrix}$$

#### Example.

To solve our example from before

$$f(x, y) = x^{3} - y + \frac{1}{4} = 0$$
$$g(x, y) = x^{2} + y^{2} - 1 = 0$$

we need its Jacobian

$$J(x,y) = \begin{pmatrix} 3x^2 & -1 \\ 2x & 2y \end{pmatrix}$$



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## Further exercises for you to try

- Search for the solution of the Example in the third quadrant by changing the initial value  $x^{(0)}$ .
- Apply Newton's method to the system

$$xe^{y} = 1$$
$$-x^{2} + y = 1$$

using  $x^{(0)} = y^{(0)} = 0$ .

#### Error analysis for multivariate Newton's method

Error and convergence analysis for the multivariate case is beyond the scope of this lecture.

**Summary.** If *f* is sufficiently differentiable, and there is a solution  $x^*$  of the system f(x) = 0 with  $J(x^*)$  nonsingular, then the Newton iterations will converge quadratically towards  $x^*$  for all  $x^{(0)}$  sufficiently close to  $x^*$ .

### **Outlook. Multivariate Newton's method**

- finding solutions in the multivariate case is hard
- especially choosing a good starting point  $x^{(0)}$  is
- systems of equations usually have multiple solutions
- how do you find the one you want?
- ▶ for large *n*: Evaluating the Jacobian  $J(\mathbf{x}^{(k)})$  is expensive
- there exist efficient versions including slow, but robust algorithms to find x<sup>(0)</sup>.