

## Lecture notes from the 23. December 2010

Please not that in the lecture it is section number 5 and not 0.

### Bisection method

Let  $f : [a, b] \rightarrow \mathbb{R}$  be continuous and with  $f(a)f(b) < 0$ .

Find  $x^* \in [a, b]$  such that  $f(x^*) = 0$ .

intermediate value theorem  $\Rightarrow$  There exists at least one root of  $f$  in  $(a, b)$ .

### Algorithm 0.1 (Bisection)

$$x_p^{(0)} := \begin{cases} a & \text{if } f(a) > 0 \\ b & \text{if } f(b) > 0 \text{ (i.e., otherwise)} \end{cases}$$

and

$$x_n^{(0)} := \begin{cases} a & \text{if } f(a) < 0 \\ b & \text{if } f(b) < 0 \text{ (i.e., otherwise)} \end{cases}$$

for  $k = 0, \dots$

$$x_m^{(k)} := \frac{x_p^{(k)} + x_n^{(k)}}{2} \quad (\text{midpoint})$$

$$x_p^{(k+1)} := \begin{cases} x_m^{(k)} & \text{if } f(x_m^{(k)}) > 0 \\ x_p^{(k)} & \text{otherwise} \end{cases}$$

$$x_n^{(k+1)} := \begin{cases} x_m^{(k)} & \text{if } f(x_m^{(k)}) < 0 \\ x_n^{(k)} & \text{otherwise} \end{cases}$$

For  $k \geq 0$  we have

1.  $|x_p^{(k)} - x_n^{(k)}| = 2^{-k}(b - a) \Rightarrow$  the length of the intervals converges to zero without loss of generality, we may assume  $f(x_m^{(0)}) > 0$  which gives

$$\begin{aligned} |x_p^{(1)} - x_n^{(1)}| &= |x_m^{(0)} - x_n^{(0)}| = \left| \frac{x_p^{(0)} + x_n^{(0)}}{2} - x_n^{(0)} \right| \\ &= \left| \frac{x_p^{(0)} - x_n^{(0)}}{2} \right| \\ &= 2^{-1}(b - a) \end{aligned}$$

Now we make an induction step  $k \rightarrow k + 1$ , assuming again without restriction  $f(x_m^{(k)}) > 0$

$$|x_p^{(k+1)} - x_n^{(k+1)}| = |x_m^{(k)} - x_n^{(k)}| = \left| \frac{x_p^{(k)} + x_n^{(k)}}{2} - x_n^{(k)} \right|$$

$$\begin{aligned}
&= \left| \frac{x_p^{(k)} - x_n^{(k)}}{2} \right| \\
&= 2^{-1} |x_p^{(k)} - x_n^{(k)}| \\
&= 2^{-1} 2^{-k} (b - a) \\
&= 2^{-(k+1)} (b - a)
\end{aligned}$$

2. We assume without limitation  $x_n^{(l)} < x_p^{(l)}$  for all  $l$ . Then

$$[x_n^{(k+1)}, x_p^{(k+1)}] \subset [x_n^{(k)}, x_p^{(k)}]$$

3.  $f(x_p^{(k)}) \geq 0$ ,  $f(x_n^{(k)}) \leq 0$  for all  $k \geq 0$

### Fixed point iteration

Remember: Find  $x^* \in \mathbb{R}$  with  $g(x^*) = x^*$

#### Algorithm 0.2 (Fixed point iteration)

Choose an initial guess  $x^{(0)}$ . For  $k \geq 0$  compute

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

Question: When does the fixed point iteration converge?

### Newton iteration

Find  $x^*$  such that  $f(x^*) = 0$ .

#### Algorithm 0.3 (Newton iteration)

Choose an initial guess  $x^{(0)}$ . For  $k \geq 0$  compute

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

The convergence of the Newton iteration depends on the initial value.

### Secant method

Find  $x^*$  such that  $f(x^*) = 0$ .

#### Algorithm 0.4 (Secant method)

Let  $x^{(0)}$  and  $x^{(1)}$  be given. For  $k \geq 1$  compute

$$x^{(k+1)} := x^{(k)} - f(x^{(k)}) \underbrace{\left( \frac{x^{(k)} - x^{(k-1)}}{f(x^{(k)}) - f(x^{(k-1)})} \right)}_{\text{inverse of the slope of the secant}}$$

## 0.1 Banach Fixed Point Theorem

### Definition 0.1

A function  $g : D \subset \mathbb{R}^n \rightarrow \mathbb{R}^n$  is called **contractive** in  $G$  with respect to the norm  $\|\cdot\|$  if there exists a constant  $q \in (0, 1)$ , i.e.,  $0 < q < 1$ , such that

$$\|g(x) - g(y)\| \leq q\|x - y\| \quad \forall x, y \in D. \quad (0.0)$$

The constant  $q$  is called **Lipschitz constant** of  $g$  in  $D$ . A function  $g$  which satisfies (0.0), where the constant  $q > 0$  may not be bounded from above by 1, is called **Lipschitz continuous**.

### Theorem 0.1

Let  $D \subset \mathbb{R}^n$  be a convex set and the function  $g : D \rightarrow \mathbb{R}^n$  differentiable. Furthermore, let  $q := \sup_{x \in D} \|Dg(x)\| < 1$ , where  $Dg(x)$  is the Jacobi matrix

$$Dg(x) = \begin{pmatrix} \frac{\partial g_1(x)}{\partial x_1} & \cdots & \frac{\partial g_1(x)}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial g_n(x)}{\partial x_1} & \cdots & \frac{\partial g_n(x)}{\partial x_n} \end{pmatrix} \quad \text{of } g(x) = \begin{pmatrix} g_1 \\ \vdots \\ g_n \end{pmatrix}.$$

Then  $g$  is a contractive function.

**Proof:** Remember:  $D \subset \mathbb{R}^n$  is convex means

$$\forall a, b \in D \Rightarrow x = ta + (1-t)b \in D \quad \forall t \in [0, 1]$$

Assume:  $x, y \in D, t \in [0, 1]$

Define:  $f(t) := g(tx + (1-t)y)$

Chain rule:  $\Rightarrow Df(t) = Dg(tx + (1-t)y)(x - y)$

$$\begin{aligned} \|g(x) - g(y)\| = \|f(1) - f(0)\| &= \left\| \int_0^1 Df(t) dt \right\| \\ &\leq \sup_{t \in [0, 1]} \|Df(t)\| \\ &= \sup_{t \in [0, 1]} \|Dg(tx + (1-t)y)(x - y)\| \\ &\leq \left( \sup_{t \in [0, 1]} \|Dg(tx + (1-t)y)\| \right) \|x - y\| \\ &= \underbrace{\left( \sup_{z \in D} \|Dg(z)\| \right)}_{< 1} \|x - y\| \end{aligned}$$

$\Rightarrow g$  is contractive. □

**Definition 0.2 (Cauchy sequence)**

A sequence  $(a_n)_{n \in \mathbb{N}}$  is called a **Cauchy sequence** or **Cauchy convergent** if and only if ( $:\Leftrightarrow$ )

$$\begin{aligned} \forall \varepsilon > 0 \exists N \in \mathbb{N} \forall n \in \mathbb{N}, n \geq N \forall k \in \mathbb{N} & : \|a_n - a_{n+k}\| < \varepsilon \\ \forall \varepsilon > 0 \exists N \in \mathbb{N} \forall n, m \in \mathbb{N}, n, m \geq N & : \|a_n - a_m\| < \varepsilon \end{aligned}$$

**Theorem 0.2**

For a sequence  $(a_n)_{n \in \mathbb{N}}$  in a complete space as  $\mathbb{R}$  the following two statements are equivalent

1.  $(a_n)_{n \in \mathbb{N}}$  is convergent.
2.  $(a_n)_{n \in \mathbb{N}}$  is a Cauchy sequence.