

# Lecture 3

## Bisection, Newton's Method, Secant Method

### Bisection

#### The Problem

Suppose we have a continuous function  $f$  on some domain  $[a, b]$ . Find  $x^* \in [a, b]$  such that

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

**Theorem 1 (Intermediate Value Theorem)** Suppose we have a continuous function  $f$  on some domain  $[a, b]$ . Then if  $k$  is some number between  $f(a)$  and  $f(b)$  then there exists at least one number  $c$  in the interval  $[a, b]$  such that  $f(c) = k$ . That is,

$$\begin{aligned} f(a) < k < f(b) \quad \text{or} \quad f(a) > k > f(b), \\ \rightarrow \quad \exists c \in [a, b] \quad \text{s.t.} \quad f(c) = k \end{aligned}$$

**Note:** The Intermediate Value Theorem can be applied at  $k = 0$  to find where  $f(c) = 0$  when  $f(a)f(b) < 0$ .

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**Algorithm 1** Bisection.

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**Require:**  $f \in \{f : \mathbb{F} \rightarrow \mathbb{F}\}$ ,  $a \in \mathbb{R}$ ,  $b \in \mathbb{R}$ ,  $k_{\max} \in \mathbb{Z}^+$ ,  $\epsilon_x \in \mathbb{R}$ ,  $\epsilon_f \in \mathbb{R}$ .

$a \leftarrow a_0$

$b \leftarrow b_0$

$k \leftarrow 1$

**while**  $k \leq k_{\max}$  **do**

$c_k \leftarrow \frac{1}{2}(a_{k-1} + b_{k-1})$

$f_k \leftarrow f(c_k)$

**if**  $f_k f(a_{k-1}) > 0$  **then**

$a_k \leftarrow c_k$

$b_k \leftarrow b_{k-1}$

**else**

$a_k \leftarrow a_{k-1}$

$b_k \leftarrow c_k$

**end if**

**if**  $|b_k - a_k| < \epsilon_x$  **or**  $|f(c_k)| < \epsilon_f$  **then**

$k \leftarrow k_{\max} + 1$

**end if**

$k \leftarrow k + 1$

**end while**

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# Newton's Method

**Theorem 2 (Newton's Method)** Given an iterate  $x^{(k)}$  approximating a zero of  $f$ , the next iterate is:

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

- Iterative procedure to locate zeros of  $f$ .
- Requires initial iterate  $x^{(0)}$  to start.
- Near true zero  $x^*$  of  $f$ , iteration converges quickly.

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**Algorithm 2** Newton's Method

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**Require:**  $f, f', x^{(0)}$ .

$k \leftarrow 0$

**while**  $k = 0, 1, 2, \dots$  **until** convergence **do**

$r^{(k)} \leftarrow f(x^{(k)})$

$\delta x^{(k)} \leftarrow -[f'(x^{(k)})]^{-1} r^{(k)}$

$x^{(k+1)} \leftarrow x^{(k)} + \delta x^{(k)}$

Test for convergence

**end while**

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▷ (evaluate nonlinear residual)

▷ (compute Newton step)

▷ (compute next iterate)

▷ (break if necessary)

**Definition 1** The **residual**  $r^{(k)}$  is defined by:

$$r^{(k)} := f(x^{(k)})$$

**Definition 2** The **Newton step**  $\delta x^{(k)}$  is defined by:

$$\delta x^{(k)} := -[f'(x^{(k)})]^{-1} r^{(k)}$$

**Example 1** Carry out Newton's method starting from  $x^{(0)} = 0.75$  to find  $x^{(3)}$  that approximates the real solution of  $x = \cos x$ .

**Example 2** Carry out Newton's method starting from  $x^{(0)} = 0.5$  to find  $x^{(3)}$  that approximates a zero of the equation  $xe^x = 2$ .

Now we have to answer these three essential questions:

### 1. Under what conditions does the algorithm converge?

Bisection converges to some  $x^*$  such that  $f(x^*) = 0$  in  $[a, b]$  if  $f$  is continuous and  $f(a)f(b) < 0$ . If there are two or more solutions, we don't know to which one it will converge. Newton iteration converges if  $x_0$  is *sufficiently close* to  $x^*$ . Usually, we do not know a priori how close is close enough and we must resort to trial and error.

### 2. How accurate will the result be?

Both methods can give us  $x^*$  up to machine precision.

### 3. How fast does it converge?

In bisection, the error  $|x^* - x^{(k)}|$  decreases by a factor of  $\frac{1}{2}$  in each iteration. In Newton iterations, the error is approximately squared in each iteration (provided it is small enough).

$$\epsilon_0, \frac{\epsilon_0}{2}, \frac{\epsilon_0}{4}, \frac{\epsilon_0}{8}, \dots \quad \text{vs.} \quad \epsilon_0, \epsilon_0^2, \epsilon_0^4, \epsilon_0^8, \dots$$

Newton's method converges very quickly, but requires the computation of  $f'(x)$ . Sometimes, we cannot compute it, for instance if  $f$  is shorthand for some complicated procedure. In that case we have two options: (1) bisection or (2) the secant method.

**Remark 1** Suppose we have two initial points,  $x_0$  and  $x_1$ . Then we can estimate the derivative of  $f$  at  $x_1$  as

$$f'(x_1) \approx \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

and substitute this in the Newton iteration:

$$x_2 = x_1 - \frac{f(x_1)}{f'(x_1)} \approx x_1 - \frac{f(x_1)(x_1 - x_0)}{f(x_1) - f(x_0)}$$

**Theorem 3 (Secant Method)** Given iterates  $x^{(k)}$  and  $x^{(k-1)}$  approximating a zero of  $f$ , compute

$$x^{(k+1)} = x^{(k)} - \frac{f(x^{(k)})(x^{(k)} - x^{(k-1)})}{f(x^{(k)}) - f(x^{(k-1)})} \quad (k \geq q)$$

Secant methods need two initial guesses:  $x^{(0)}$  and  $x^{(1)}$ .

**Remark 2** Some remarks on secant methods:

- This method uses a *finite difference* approximation to  $f'$ .
- Asymptotically (meaning if  $|x^{(k)} - x^*|$  is small enough) the secant method converges as fast as Newton's method does.
- The secant method has extensions to problems with more than 1 unknown, but in this case Newton's method tends to be less cumbersome.
- The secant method is a *second order recurrence relation*. It relates the next approximation to the two previous approximations.
- If we can find an  $a$  and  $b$  such that  $x^* \in [a, b]$ , then  $x_0 = a$  and  $x_1 = b$  is a good starting point.

# Recursion

Recurrence and iteration really mean procedures in which we repeat the same action over and over. One way to program this is by using `for` and `while` loops. We can also make the recurrent nature of the computation explicit by making the function call itself. This is called recursive programming. See the Python code below for a simple example of recursion to calculate the factorial function:

```
def fact(k):  
    if k == 1:  
        return 1  
    else:  
        return fact(k-1) * k
```