### MA 204 Numerical Methods

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#### **Contents**

 Solution of a nonlinear equation, bisection and secant methods, Newton's method, rate of convergence, solution of a system of nonlinear equations.

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- Solution of a nonlinear equation, bisection and secant methods, Newton's method, rate of convergence, solution of a system of nonlinear equations.
- Interpolation by polynomials, divided differences, error of the interpolating polynomial, piecewise linear and cubic spline interpolation.

### Motivation

#### Nonlinearity

One of the most frequent problem in engineering and science is to find the  $\mathsf{root}(\mathsf{s})$  of a non-linear equation

$$f(x) = 0. (1)$$

Here,

- $f:[a,b] \to \mathbb{R}$  is a nonlinear function in x;
- $f \in C^1[a, b]$ ;
- Roots are isolated.

## Root of the equation

#### Definition 1

Given a nonlinear function  $f:[a,b]\to\mathbb{R}$ , find a value of r for which f(r)=0. Such a solution value for r is called a **root** of the equation

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### Approximation to a root

A point  $x^* \in \mathbb{R}$  such that

- $|r-x^*|$  is **very small**, and
- $f(x^*)$  is **very close** to 0.

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**Advantages:** Always converges

**Disadvantages:** Locating the root initially

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- 2 Closed Domain Methods (Non- Bracketing Methods)
  - Secant method
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  - Newton's method (Newton Raphson method)

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- 2 Closed Domain Methods (Non- Bracketing Methods)
  - Secant method
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  - Newton's method (Newton Raphson method)

Advantages: No need to locate the root initially

Disadvantages: May not converge

■ Description of the method/ Basic idea

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- Error analysis of the iteration and convergence

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- Application and example

This method is based on Mean Value Theorem.

Assume that two initial guesses to  $\alpha$  are known. Let these be  $x_0$  and  $x_1$ . They may occur on opposite side of  $\alpha$  or on the same side of  $\alpha$ .

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- Equation of the secant line

$$y - f(x_1) = \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x - x_1)$$
 (2)

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$$0 - f(x_1) = \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x_2 - x_1)$$

$$\implies x_2 = x_1 - \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x_2 - x_1).$$
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$$\Longrightarrow x_2 = x_1 - \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x_2 - x_1).$$
(3)

■ Having found  $x_2$ , we can drop  $x_0$  and use  $x_1, x_2$  as a new set of approximate value for  $\alpha$ . This leads to an improved value  $x_3$ ; and this process can be continued indefinitely.

■ The general iteration formula for the secant method is

$$x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \quad n \ge 1.$$
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- It is called a two-point method, since two approximation values are needed to obtain an improved value.
- The Bisection method is also a two-point method, but the Secant method will almost always converge faster than bisection.

Note that only one function evaluation is needed per step for the Secant method after  $x_2$  has been determined. In contrast, each step of Newton's method requires an evaluation of both

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- Newton's method or the Secant method is often used to refine an answer obtained by another technique, such as the bisection method, since these methods requires good first approximation but generally give rapid convergence.

# How fast is the convergence?

#### Definition 2

Let  $\{x_n\}_{n\geq 1}$  be a sequence that converges to  $\alpha$ . If positive constants  $\lambda$  and p exist with

$$\lim_{n\to\infty}\frac{|x_{n+1}-\alpha|}{|x_n-\alpha|^p}=\lambda,$$

then  $\{x_n\}_{n\geq 1}$  is said to converge to  $\alpha$  of order p, with assymptotic error constant  $\lambda$ . If p=1, the method is called linear. If p=2, the method is called quadratic.

# How fast is the convergence?

Can we find the exponent p such that

$$|x_{n+1} - \alpha| \approx |x_n - \alpha|^p? \tag{5}$$

■ Answer:  $p = \frac{1}{2}(1 + \sqrt{5}) \approx 1.62$ . This is called **super linear** convergence (1 .

■ The general iteration formula for the secant method is

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■ Let  $\varepsilon_n = x_n - \alpha$  and  $\varepsilon_{n+1} = x_{n+1} - \alpha$ . So,  $\varepsilon_n$  and  $\varepsilon_{n+1}$  denote the errors in the root at the nth and (n+1)th iterations.

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- **■** (6) ⇒

$$\varepsilon_{n+1} = \varepsilon_n - \frac{f(\alpha + \varepsilon_n)(\varepsilon_{n+1} - \varepsilon_n)}{f(\alpha + \varepsilon_n) - f(\alpha + \varepsilon_{n-1})}$$
(7)

Assume that f is twice differentiable and  $f'(\alpha), f''(\alpha) \neq 0$ . By Taylor's formula (with very small  $\varepsilon$ )

$$f(\alpha + \varepsilon) = f(\alpha) + \varepsilon f'(\alpha) + \frac{\varepsilon^2}{2} f''(\alpha) + R_2(\varepsilon).$$

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Now,  $f(\alpha)=0$ , and  $\varepsilon$  is very small, and  $R_2(\varepsilon)$  is the remainder term.  $R_2(\varepsilon)$  vanishes as  $\varepsilon\to 0$  at a faster rate than  $\varepsilon^2$ . Therefore,

$$f(\alpha + \varepsilon) \approx \varepsilon f'(\alpha) + \frac{\varepsilon^2}{2} f''(\alpha).$$

Let 
$$M = \frac{f''(\alpha)}{2f'(\alpha)}$$
.

 $f(\alpha + \varepsilon_n) \approx \varepsilon_n f'(\alpha) \Big( 1 + \varepsilon_n M \Big).$ 

$$\varepsilon_{n+1} \approx \frac{f''(\alpha)}{2f'(\alpha)} \varepsilon_{n-1} \varepsilon_n$$
(8)

■ (8) tells us that, as  $n \to \infty$ , the error tends to zero faster than a linear function but not quadratically!

#### **Error Estimate**

By Mean Value Theorem,

$$f(\alpha) - f(x_n) = f'(c_n)(\alpha - x_n)$$
 (9)

where  $c_n$  lies between  $x_n$  and  $\alpha$ . So, if  $x_n \to \alpha$ , then  $c_n \approx x_n$  for large n, and we have

$$\alpha - x_n \approx -\frac{f(x_n)}{f'(c_n)}$$

$$\approx -\frac{f(x_n)}{f'(x_n)}$$

$$\approx -f(x_n)\frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}$$

$$\approx x_{n+1} - x_n.$$

Thus,  $\alpha - x_n \approx x_{n+1} - x_n$ .