

# COMPUTER SCIENCE 349A

## Handout Number 11

### ORDER OF CONVERGENCE OF THE SECANT METHOD AND BISECTION METHOD

The Secant method iterative formula is

$$x_{i+1} = x_i - f(x_i) \frac{x_i - x_{i-1}}{f(x_i) - f(x_{i-1})}.$$

Therefore, if  $x_t$  denotes an exact zero of  $f(x)$ ,

$$\begin{aligned} x_{i+1} - x_t &= (x_i - x_t) - f(x_i) \frac{x_i - x_{i-1}}{f(x_i) - f(x_{i-1})} \\ &= (x_i - x_t)(x_{i-1} - x_t) \left[ \frac{1}{\frac{f(x_i) - f(x_{i-1})}{x_i - x_{i-1}}} \right] \left[ \frac{\frac{f(x_i) - f(x_{i-1})}{x_i - x_{i-1}} - \frac{f(x_i) - f(x_t)}{x_i - x_t}}{x_{i-1} - x_t} \right] \end{aligned}$$

By the Mean Value Theorem, the first large square bracket term above is equal to  $\frac{1}{f'(\xi_i)}$ , for some value  $\xi_i$  between  $x_i$  and  $x_{i-1}$ .

The term in the second large square bracket above is equal to  $\frac{f''(\eta_i)}{2}$ , for some value  $\eta_i$  in an interval spanned by  $x_i, x_{i-1}$  and  $x_t$  as this is a “second divided difference” -- see page 97 of the 7<sup>th</sup> ed. (page 94 of the 6<sup>th</sup> ed.).

Thus, with the usual notation that  $E_k = x_k - x_t$  denotes the error of the  $k$ -th approximation  $x_k$ , the above can be written as

$$E_{i+1} = E_i E_{i-1} \left[ \frac{1}{f'(\xi_i)} \right] \left[ \frac{f''(\eta_i)}{2} \right]$$

or

$$\frac{E_{i+1}}{E_i E_{i-1}} = \frac{f''(\eta_i)}{2f'(\xi_i)}.$$

Thus

$$(1) \quad \lim_{i \rightarrow \infty} \left| \frac{E_{i+1}}{E_i E_{i-1}} \right| = \lim_{i \rightarrow \infty} \left| \frac{f''(\eta_i)}{2f'(\xi_i)} \right| = \left| \frac{f''(x_t)}{2f'(x_t)} \right|.$$

This gives a relationship **between 3 successive errors** of the computed approximations using the Secant method. However, this does not indicate the **order**  $\alpha$  of the Secant method, which requires that the errors of 2 successive approximations be related by

$$(2) \quad \lim_{i \rightarrow \infty} \frac{|E_{i+1}|}{|E_i|^\alpha} = \lambda, \text{ for some constant } \lambda.$$

The following analysis shows how to rewrite (1) in a form (2). From (1), it follows that for sufficiently large values of  $i$ ,

$$(3) \quad |E_{i+1}| \approx C |E_i| |E_{i-1}|, \text{ for some constant } C.$$

From (2), it follows that for sufficiently large values of  $i$ ,

$$(4) \quad |E_{i+1}| \approx \lambda |E_i|^\alpha \text{ and } |E_i| \approx \lambda |E_{i-1}|^\alpha.$$

Substituting (4) into (3) gives

$$\lambda |E_i|^\alpha \approx C |E_i| \left[ \frac{|E_i|}{\lambda} \right]^\frac{1}{\alpha},$$

which implies that

$$\lambda^{1+\frac{1}{\alpha}} |E_i|^\alpha \approx C |E_i|^{1+\frac{1}{\alpha}}.$$

From this it follows that

$$\alpha = 1 + \frac{1}{\alpha} \Rightarrow \alpha^2 - \alpha - 1 = 0 \Rightarrow \alpha = \frac{1 + \sqrt{5}}{2} \approx 1.618,$$

which is the **order of the Secant method**.

**Note:** this value  $\alpha$  is known as the “golden ratio”, and occurs in many diverse applications.

## ORDER OF CONVERGENCE OF THE BISECTION METHOD

An alternate definition of **linear convergence** (similar to the definition given in class except that it does not involve a limit) is

$$|E_i| \leq c |E_{i-1}| \quad \text{or} \quad |x_t - x_i| \leq c |x_t - x_{i-1}|$$

for some constant  $c$  such that  $0 < c < 1$ .

Applying this inequality recursively gives

$$|x_t - x_i| \leq c^i |x_t - x_0|.$$

From Page 2 of Handout Number 8, for the Bisection Method we had

$$|x_t - x_i| \leq \left(\frac{1}{2}\right)^i \Delta x^0, \quad \text{where } \Delta x^0 = x_u - x_\ell$$

and  $[x_\ell, x_u]$  is the initial interval. This implies linear convergence with the above definition, and  $c = \frac{1}{2}$ .