

## 2-3 Zeros of a Function

노트 제목

①

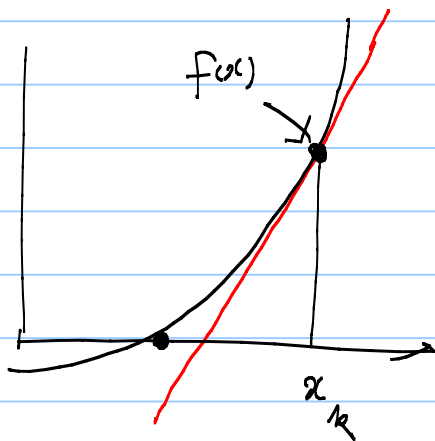
Recall • FONC:  $h'(x^*) = 0$

• To find  $x^* \Leftrightarrow$  to find sol. of  $f(x) = 0$   
( $f(x) = h'(x)$ )

• objective: to study numerical method to find solution of  $f(x) = 0$ .

- i) Use an one point to start  
→ Newton's method, Secant Method
- ii) Use two points  $\Leftrightarrow$  one pt + search Interval  
→ Bisection method, etc

### ◀ Newton's Method ▶



- i) Start with  $x_k$   
( $x_k$ :  $k$ th guess,  $k=1,2,\dots$ )
- ii) Use linear approximation of  $f(x)$  using  $f(x_k)$  &  $f'(x_k)$
- iii) Solve for an approx sol
- iv) repeat until convergence

(2)

Analysis

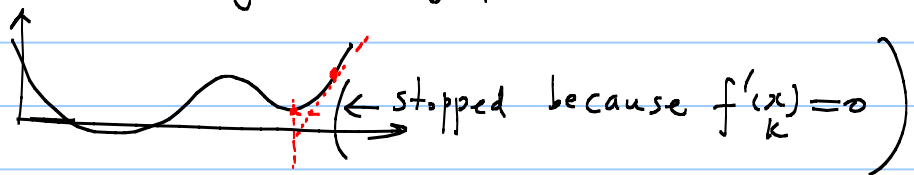
a)  $\bar{f}(x) = f(x_k) + f'(x_k)(x - x_k)$   
around  $x_k$

b)  $\bar{f}(x) \equiv 0 \rightarrow x = x_{k+1}$

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} \quad - (1)$$

Remark on Newton's Method

i) Newton's Method: Not guaranteed to work for arbitrary starting points



ii) Convergence property?

Convergence Condition:  $\left| \frac{f(x)f''(x)}{[f'(x)]^2} \right| < 1 \quad (2)$

③

★  $\square$  Rate of Convergence : 2nd-Order

$$\text{i.e.; } |e_{k+1}| = K |e_k|^2 \quad (3)$$

$\uparrow$  error at (k+1)th iteration       $\nwarrow$  error at the kth iteration

$\square$  Convergence Analysis  
(to show Eqs (2) and (3))

▪ Trick: Rewrite  $f(x) = 0$  as

$$\rightarrow \frac{f(x)}{f'(x)} = 0 \quad (\text{if } f'(x) \neq 0)$$

$$\rightarrow \boxed{x = x - \frac{f(x)}{f'(x)} \equiv g(x)} \quad (4)$$

$\nwarrow$  This form appeared in Eq. (1)

$\square$  Observation when Eq. (4) is used

○ if  $x^*$  satisfies  $f(x^*) = 0$ ,  
then  $x^* = g(x^*)$  holds

④

② Newton's Algorithm can be viewed as

$$x_{k+1} = g_{1k}(x_k) \quad (k=1, 2, \dots)$$

③ Define error  $e_k$

$$e_k = x_k - x^* \quad (\text{if we happen to know } x^*)$$

A) To find the convergence condition (2):

$$e_{k+1} \stackrel{\textcircled{3}}{=} x_{k+1} - x^*$$

$\uparrow$   $\uparrow$   
 $g(x_k)$   $g(x^*)$  (Exact relation)  
by Newton's Algorithm

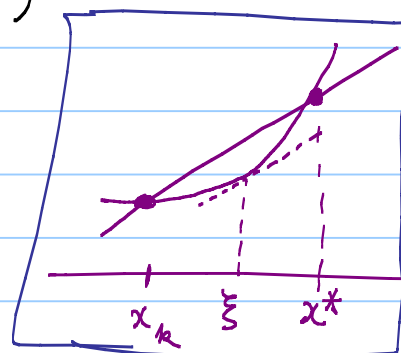
$$= g(x_k) - g(x^*)$$

$$= \frac{g(x_k) - g(x^*)}{x_k - x^*} (x_k - x^*)$$

By Mean Value Theorem

$$= g'(\xi) (x_k - x^*)$$

$$= g'(\xi) e_k \quad (5)$$



(5)

(Mean value Theorem: When  $g(x)$  is continuous on  $[a, b]$ , then at some point  $\xi \in [a, b]$

$$g'(\xi) = \frac{g(b) - g(a)}{b - a}$$

provided that  $g'(x)$  exists at all interior points.)

$$\therefore |e_{k+1}| = |g'(\xi)| |e_k| \quad (6)$$

Observation: For  $|e_{k+1}|$  to be less than  $|e_k|$ , the following eq must hold

$$|g'(\xi)| \leq K < 1 \quad (7)$$

$$\text{i.e.,} \quad |g'(\xi)| \stackrel{(4)}{=} \left| \frac{d}{d\xi} \left( \xi - \frac{f(\xi)}{f'(\xi)} \right) \right|$$

$$= \left| 1 - \frac{f'(\xi)f'(\xi) - f(\xi)f''(\xi)}{[f'(\xi)]^2} \right|$$

$$= \left| \frac{f(\xi)f''(\xi)}{[f'(\xi)]^2} \right| < 1$$

condition for convergence

$\Rightarrow$  Equation (2)



⑦

$$= \frac{1}{2} g''(\xi) e_k^2$$

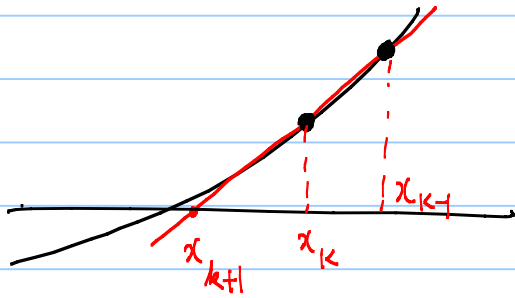
$$\therefore e_{k+1} = \beta e_k^2$$
$$|e_{k+1}| = |\beta| |e_k|^2 \quad \left( \overset{\text{2nd-order}}{\left( 2 \right)} \right)$$

$\Rightarrow$  "The Rate of Convergence is of Order 2."

\* Remark: For the case of double roots, the convergence rate becomes quadratic if  $x_{k+1} = x_k - 2 \frac{f(x)}{f'(x)}$  is used.

## « Secant Method »

A straight line that cuts a curve at one or more points (In this case, 2 points)  $\textcircled{P}$



① Similar to Newton's method

② approximate  $f'(x_k)$  as

$$f'(x_k) = \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}}$$

→ No derivative of  $f(x)$  is need  
(Useful when  $df/dx$  is difficult to calculated analytically)

$$\therefore f(x) = f(x_k) + f'(x_k)(x - x_k) + \dots$$

$$\Rightarrow \bar{f}(x) = f(x_k) + \frac{f(x_k) - f(x_{k-1})}{x_k - x_{k-1}} (x - x_k) + \dots$$

$\Delta f_k$

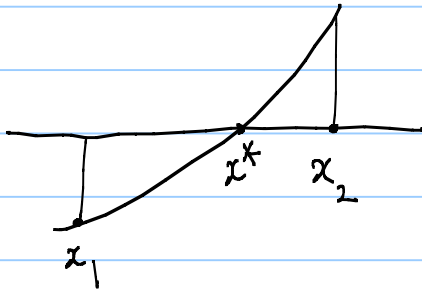
Imposing  $\bar{f}(x) = 0$ ,  $x_{k+1}$  is obtained

$$x_{k+1} = x_k - \frac{x_k - x_{k-1}}{f_k - f_{k-1}} f_k$$



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## « Bi-section Method »

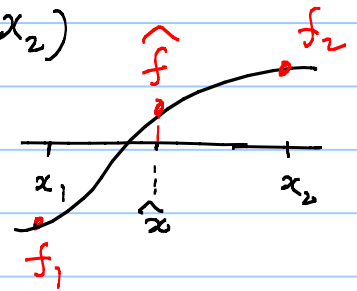


• Start with  $I = [x_1, x_2] \ni x^*$

i)  $\hat{x} = \frac{1}{2}(x_1 + x_2)$

ii) if  $f_1 \hat{f} < 0$

$$I_{\text{New}} = [x_1, \hat{x}]$$



else

$$I_{\text{New}} = [\hat{x}, x_2]$$

ii) Repeat ii) until convergence

Remark: Bisection Methods can be Combined with Polynomial approaches,