

(proof of convergence of bisection method; slides P.6)  
Chop the task into pieces.

①  $\lim a_n, \lim b_n, \lim c_n$  exist and they all the same.

② Call the limit,  $\xi$ , then  $f(\xi) = 0$ .

③  $|c_n - \xi| < 2^{-(n+1)}(b-a)$

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① Observe  $a_0 \leq a_1 \leq a_2 \leq \dots \leq b$  by construction. Therefore  $\lim a_n$  exists.

/x Math 3B - monotone sequence theorem.

If  $\{a_n\}$  is nondecreasing (or nonincreasing) and bounded above (or bounded below), the limit exists. [This is "half-version"]

If  $\{a_n\}$  is monotonic (i.e., only nonincreasing or only nondecreasing) and bounded (i.e., bounded from above and below), it converges. [This is "two-sided-ver"] \*/

Likewise  $b_0 \geq b_1 \geq b_2 \geq \dots \geq a$ . Therefore  $b_n$  also converges.

Let  $\lim a_n = \xi_1$ , and  $\lim b_n = \xi_2$ .

We know the length of  $[a_n, b_n]$  gets halved from the construction. Therefore,  $b_n - a_n \rightarrow 0$  as  $n \rightarrow \infty$ . Then, we must have

$$\begin{aligned} 0 &= \lim (b_n - a_n) = \lim b_n - \lim a_n \\ &= \xi_2 - \xi_1 \end{aligned}$$

$\Rightarrow \xi_1 = \xi_2$  Call this common limit  $\xi$ .

Lastly for ①, we have  $a_n \leq c_n \leq b_n$ .

Therefore, sandwich theorem says

$$\begin{array}{ccccc} \lim a_n & \leq & \lim c_n & \leq & \lim b_n \\ \downarrow & & \downarrow & & \downarrow \\ \xi & & \xi & & \xi \end{array}$$

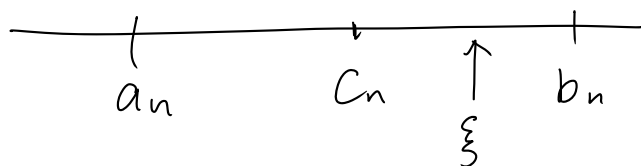
Thus,  $\lim c_n = \xi$ .

(2) Taking limit of  $f(a_n)f(b_n) \leq 0$ , we have  
 $f(\xi)f(\xi) \leq 0$ . The only possibility is  
 $f(\xi) = 0$ .

/\* Here, continuity of  $f$  is used.

$f$  conti at  $a \iff f(a_n) \rightarrow f(a) \quad \forall \{a_n\} \text{ s.t. } a_n \rightarrow a$

(3) Observe  $\xi \in [a_n, b_n] \quad \forall n$ . Thus,



$$|c_n - \xi| < \frac{1}{2} \cdot \text{length}([a_n, b_n]) \\
= \frac{1}{2} \cdot \frac{1}{2} \text{length}([a_{n-1}, b_{n-1}])$$

$$= \left(\frac{1}{2}\right)^n \text{length}([a_1, b_1])$$

$$= \left(\frac{1}{2}\right)^{n+1} \underbrace{\text{length}([a_0, b_0])}_{= b_0 - a_0 = b - a}$$

(proof of quadratic conv. of Newton's method.)

Put  $e_n = x_n - \xi$ .

Subtract  $\xi$  from the method and sneak in  $f(\xi)$

$$x_{n+1} - \xi = x_n - \xi - \frac{f(x_n) - f(\xi)}{f'(x_n)}$$

$$\begin{aligned} e_{n+1} &= \frac{1}{f'(x_n)} (f(\xi) - f(x_n) - f'(x_n)(\xi - x_n)) \\ (*) \left\{ \begin{aligned} & \left( f(\xi) = f(x_n) + f'(x_n)(\xi - x_n) + \frac{f''(c_n)}{2!}(\xi - x_n)^2, \quad c_n \in (\xi, x_n) \right. \\ & \quad \left. \text{or } (x_n, \xi) \right) \\ & = \frac{1}{f'(x_n)} \cdot \frac{f''(c_n)}{2} e_n^2 \end{aligned} \right. \quad \text{as } e_n \rightarrow 0 \end{aligned}$$

Therefore, IF  $e_n \rightarrow 0$  as  $n \rightarrow \infty$ , then

$x_n \rightarrow \xi$ , and in turn,  $f'(x_n) \rightarrow f'(\xi)$

and  $f''(c_n) \rightarrow f''(\xi)$  as  $n \rightarrow \infty$ .

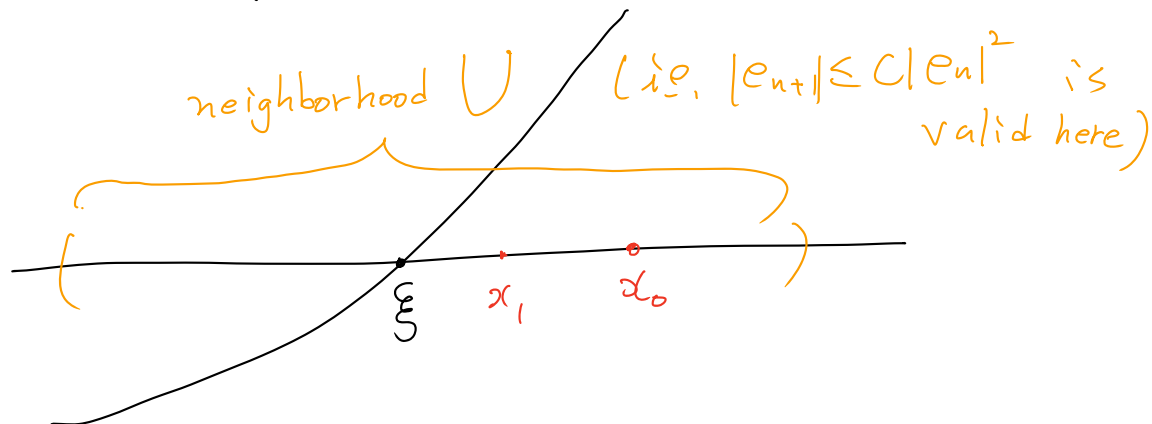
Then, dividing the error eqn (\*) by  $e_n^2$  and taking limit  $e_n \rightarrow 0$ , we have

$$\frac{e_{n+1}}{e_n^2} \rightarrow \frac{f''(\xi)}{2f'(\xi)} \text{ as } e_n \rightarrow 0. \text{ That is,}$$

$$\left| \frac{e_{n+1}}{e_n^2} \right| \leq C \text{ if } e_n \approx 0. \text{ Or roughly } |e_{n+1}| \approx C|e_n|^2$$

$\hookrightarrow$  just a generic constant as long as it is fixed.

Now, we prove the IF part.



Let  $|e_{n+1}| \leq C|e_n|^2$  on  $U$

Choose  $x_0$  so that  $|e_0| = |x_0 - \xi| < \frac{1}{2C}$   
and  $x_0 \in U$ .

/ \* This is where "if the initial guess is sufficiently close to the zero" comes in. \* /

Then, we see

$$|e_1| \leq C|e_0|^2 \leq C \cdot \frac{1}{2C} \cdot |e_0| = \frac{1}{2} |e_0|$$

Therefore,  $x_1 \in U$  too.

Repeat this so that

$$\begin{aligned} |e_2| &\leq C|e_1|^2 \leq C \cdot \left(\frac{1}{2}\right)^2 |e_0| \cdot |e_0| \\ &\leq C \cdot \left(\frac{1}{2}\right)^2 \cdot \frac{1}{2C} \cdot |e_0| = \left(\frac{1}{2}\right)^3 |e_0| \end{aligned}$$

$$\dots \quad |e_n| \leq \left(\frac{1}{2}\right)^{2^{n-1}} |e_0|$$

Thus,  $e_n \rightarrow 0$  as  $n \rightarrow \infty$ . ◻

Thm (Taylor thm)

If  $f \in C^{n+1}$  near a point  $x$  (i.e. on  $(x-\delta, x+\delta)$ ),  
then for any  $y \in (x-\delta, x+\delta)$ , we have

(Lagrange remainder ver.)

$$f(y) = f(x) + f'(x)(y-x) + \frac{f''(x)}{2!}(y-x)^2 + \dots + \frac{f^{(n)}(x)}{n!}(y-x)^n \\ + \frac{f^{(n+1)}(\xi_y)}{(n+1)!}(y-x)^{n+1} \quad (\xi_y \in (x, y) \text{ or } (y, x))$$

and

(Integral remainder ver.)

$$f(y) = f(x) + f'(x)(y-x) + \frac{f''(x)}{2!}(y-x)^2 + \dots + \frac{f^{(n)}(x)}{n!}(y-x)^n \\ + \int_x^y \frac{f^{(n+1)}(t)}{n!} \cdot (y-t)^n dt$$

(proof of Taylor theorem with Lagrange remainder)

If  $y=x$ , there is nothing to prove. So,  $y \neq x$ .

Set  $M = (f(y) - T_n(y; x)) / (y-x)^{n+1}$   
so that  $f(y) - T_n(y; x) - M(y-x)^{n+1} = 0$ .

We want: there is  $\xi_y \in (x, y)$  s.t.  $f^{(n+1)}(\xi_y) = (n+1)!M$

Introduce

$$g(t) = f(t) - T_n(t; x) - M(t-x)^{n+1}$$

Note that  $f(x) = T_n(x; x)$ ,  $f'(x) = T_n'(x; x)$ ,  
 $\dots$ ,  $f^{(n)}(x) = T_n^{(n)}(x; x)$ . Therefore,

$$g(x) = g'(x) = g''(x) = \dots = g^{(n)}(x) = 0 \quad \text{since}$$

$M(t-x)^{n+1}$  has zero at  $x$  of order  $n+1$ .

By construction of  $M$ , we have  $g(y) = 0$   
Apply MVT (mean value theorem) to  $g$ . Then,

there is  $\xi_1 \in (x, y)$  s.t.  $g'(\xi_1) = 0$ . Apply

MVT to  $g'$  on  $[x, \xi_1]$ . Then, there is

$\xi_2 \in (x, \xi_1)$  s.t.  $g''(\xi_2) = 0$ . Repeat this

so that there is  $\xi_n \in (x, \xi_{n-1})$  s.t.  $g^{(n)}(\xi_n) = 0$ .

Repeat once more to have  $\xi_{n+1} = \xi_y \in (x, \xi_n) \subset (x, y)$

s.t.  $g^{(n+1)}(\xi_y) = 0$ . That is  $f^{(n+1)}(\xi_y) = (n+1)!M$

(proof of global conv. of Newton's method  
for convex fn's)

Since  $f$  is increasing and has a  
zero by the assumption, zero is unique.

Since  $f \in C^2$  and convex,  $f''(x) \geq 0$   
 $\forall x \in \mathbb{R}$ . Also,  $f'(x) > 0 \quad \forall x \in \mathbb{R}$   
since  $f$  is increasing.

Now, recall the error equation (\*)

$$\begin{aligned} e_{n+1} &= \frac{1}{f'(x_n)} (f(\xi) - f(x_n) - f'(x_n)(\xi - x_n)) \\ &= \underbrace{\frac{1}{f'(x_n)}} \cdot \underbrace{\frac{f''(x_n)}{2}}_{\text{positive}} e_n^2 \end{aligned}$$

Deduce  $e_{n+1} = x_{n+1} - \xi \geq 0$  no matter  
 $e_n \geq 0$  or not.  $\Rightarrow e_1, e_2, e_3, \dots \geq 0$ .

This, in turn, yields  $f(x_n) \geq f(\xi) = 0$  ( $n=1,2,\dots$ )  
since  $f$  increasing:  $x_n \geq \xi \Rightarrow f(x_n) \geq f(\xi)$

$$\text{From } \underbrace{x_{n+1} - \xi}_{e_{n+1}} = \underbrace{x_n - \xi}_{e_n} - \underbrace{\frac{f(x_n)}{f'(x_n)}}_{\text{positive}},$$

We have  $e_{n+1} \leq e_n$  for  $n=1,2,\dots$



Again, monotone sequence theorem says  $\{e_n\}$  converges. But we don't know the limit. Call it  $e$ .

$e_{n+1} \leq e_n$  also implies

$x_n > x_{n+1} > \dots \geq \xi$ . The same theorem applies so that  $\lim x_n$  exists.

Call it  $z$ . Taking  $n \rightarrow \infty$  in the following

$$\underbrace{x_{n+1} - \xi}_{e_{n+1}} = \underbrace{x_n - \xi}_{e_n} - \underbrace{\frac{f(x_n)}{f'(x_n)}}_{\text{positive}}$$

$$\downarrow$$
$$\cancel{e} = \cancel{e} - \frac{f(z)}{f'(z)} \Rightarrow f(z) = 0$$

That is  $z = \xi$ , the unique zero.

(Newton's method for  $\sqrt{a}$ .)

Given  $a > 0$ ,  $\sqrt{a}$  can be defined by a positive root of

$$f(x) = x^2 - a.$$

Thus,

$$f'(x) = 2x.$$

Therefore the Newton's method reads

$$\begin{aligned} x_{n+1} &= x_n - f(x_n)/f'(x_n) \\ &= x_n - (x_n^2 - a)/2x_n \\ &= x_n - \frac{x_n}{2} + \frac{a}{2x_n} \\ &= \frac{x_n}{2} + \frac{a}{2x_n} \end{aligned}$$

proof of superlinear conv. of secant method.

Similarly to Newton's method, subtract the zero  $\xi$  from the method

$$x_{n+1} - \xi = x_n - \xi - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})} \quad (*)$$

Use MVT:  $f(x_n) - f(x_{n-1}) = f'(\eta_n)(x_n - x_{n-1})$

Plug this in, we obtain

$$\begin{aligned} e_{n+1} &= e_n - \frac{f(x_n)}{f'(\eta_n)} \\ &= \underbrace{f(\xi) + f'(\xi)(\xi - x_n)}_{=0} + \frac{f''(\alpha_n)}{2} (\xi - x_n)^2 \end{aligned}$$

$$= e_n \left( 1 - \underbrace{\frac{f'(\xi)}{f'(\eta_n)}}_{\rightarrow 1} + \underbrace{\frac{f''(\alpha_n)}{2} (\xi - x_n)}_{\rightarrow 0} \right)$$

$\rightarrow 0$

We can still extract something tending to 0!

This time we continue (\*) ±ξ → e<sub>n</sub> - e<sub>n-1</sub>

$$e_{n+1} = \frac{(f(x_n) - f(x_{n-1}))e_n - f(x_n) \overbrace{(x_n - x_{n-1})}}{f(x_n) - f(x_{n-1})}$$

$$= \frac{f(x_n)e_{n-1} - f(x_{n-1})e_n}{f(x_n) - f(x_{n-1})} \approx \frac{f'(x_n) - f'(x_{n-1})}{f'(x_n) - f'(x_{n-1})} \quad \text{("see below")}$$

$$= e_n e_{n-1} \frac{\frac{f(x_n)}{e_n} - \frac{f(x_{n-1})}{e_{n-1}}}{f(x_n) - f(x_{n-1})}$$

$$= e_n e_{n-1} \frac{(x_n - x_{n-1})}{f(x_n) - f(x_{n-1})} \frac{f''(\xi) \cdot (O(e_n^2) + O(e_{n-1}^2))}{f'(x_n) - f'(x_{n-1})} \rightarrow C \text{ as } e_n \rightarrow 0$$

$$= e_n e_{n-1} \frac{f''(\xi)}{f'(\xi)} \cdot (O(e_n^2) + O(e_{n-1}^2))$$

$$f(x_n) = f(\xi) + f'(\xi)(\xi - x_n) + \frac{f''(\xi)}{2}(\xi - x_n)^2 + \dots$$

$$\Rightarrow f(x_n)/e_n = -f'(\xi) - \frac{f''(\xi)}{2}(\xi - x_n) + O(e_n^2)$$

$$f(x_{n-1}) = f(\xi) + f'(\xi)(\xi - x_{n-1}) + \frac{f''(\xi)}{2}(\xi - x_{n-1})^2 + O(e_{n-1}^2)$$

$$\Rightarrow f(x_{n-1})/e_{n-1} = -f'(\xi) - \frac{f''(\xi)}{2}(\xi - x_{n-1}) + \dots$$

$$\Rightarrow f(x_n)/e_n - f(x_{n-1})/e_{n-1} = f''(\xi)(x_n - x_{n-1}) + O(e_n^2) + O(e_{n-1}^2)$$

In short,

$$e_{n+1} \approx C e_n e_{n-1} \quad \text{--- (**)}$$

Let's assume that a fixed rate of conver.  
is available and call it  $\varphi$ .

$$|e_{n+1}| \approx A |e_n|^\varphi \quad \text{when } |e_n| \text{ is small}$$

Plug this into (\*\*).

$$\begin{aligned} |e_{n+1}| &\approx C |e_n| |e_{n-1}| \\ A |e_n|^\varphi &\quad C A |e_{n-1}|^{\varphi+1} \\ A (A |e_{n-1}|^\varphi)^\varphi & \\ A^{1+\varphi} |e_{n-1}|^{\varphi^2} & \end{aligned}$$

$$\Rightarrow \quad \underbrace{|e_{n-1}|^{\varphi^2 - \varphi - 1}} \approx C A^{-\varphi} \quad (\text{constant})$$

tending to 0 or  $\infty$  unless  $\varphi^2 - \varphi - 1 = 0$

$$\Rightarrow \quad \varphi = \frac{1 \pm \sqrt{5}}{2}$$

$$\Rightarrow \quad \varphi = \frac{1 + \sqrt{5}}{2} \quad (\text{take positive one})$$