

Convergence Order

Numerical Integration methods we've seen converge as follows — under appropriate assumptions —

$$|\epsilon_{i+1}| \approx \alpha |\epsilon_i|_i \quad \left(\begin{array}{l} \text{not the mathematically} \\ \text{rigorous defn. - see} \\ \text{Ralston 78, e.g.} \end{array} \right)$$

this leads to

$$|\epsilon_i| \approx c^i$$

Let's look at another very important numerical method.

Example

Square-Root Algorithm, given N

$$x_0 = 1 ;$$

$$x_{i+1} = \frac{1}{2}(x_i + N/x_i) \text{ until } |x_{i+1} - x_i| < \epsilon.$$

Intuition: at convergence $x = \frac{1}{2}(x + N/x)$
... if convergence takes place

$$\Rightarrow x = N/x \Rightarrow x^2 = N$$

iteration	<u>$N=2$, Error</u>		<u>$N=100$, Error</u>	
1	0.08578	— — — —	40.50	
2	0.00245	— — — —	16.24	
3	0.000002112	— — — —	5.025	
4	*	— — — —	0.8404	
5		— — — —	0.03258	
6		— — — —	0.000052	
7		— — — —	*	

Empirically, we observe that

of places doubles (approximately)
every iteration — eventually

Note that this is ultimate behavior.

Eg. if N is very small

$$\chi_{i+1} \approx \chi_i/2 \dots \text{until we get close}$$

of places doubling is example of

$$\epsilon_{i+1} \approx d(\epsilon_i)^2 \rightarrow \begin{array}{l} \text{called} \\ \text{quadratic} \\ \text{convergence} \end{array}$$

$$\epsilon_{i+1} \approx d|\epsilon_i| \rightarrow \begin{array}{l} \text{called} \\ \text{linear} \\ \text{convergence} \end{array}$$

$$\text{Linear } \epsilon_{i+1} \approx c^i$$

$$\text{quadratic } \epsilon_{i+1} \approx c^{2^i}$$

this is an example of a root-finding,
or zero-finding problem in one real variable —

$$x^2 - N = 0$$

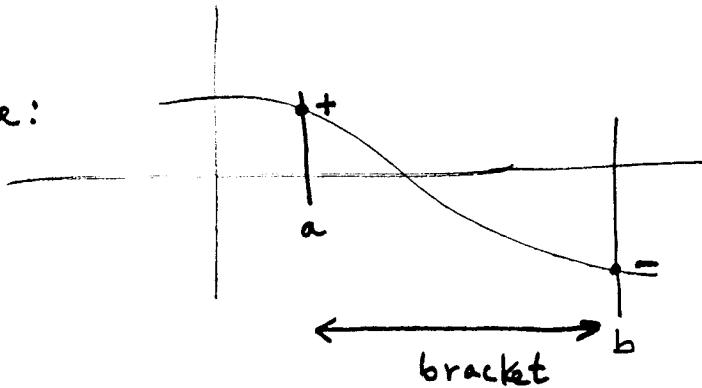
Root Finding in one-D:

2.2.3

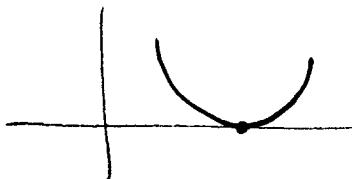
- Issues:
- choosing starting point, interval(?)
 - convergence?
 - rate of convergence?

Dangers lurk! \rightarrow insight & knowledge is crucial

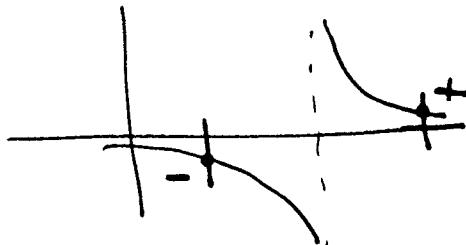
Good case:



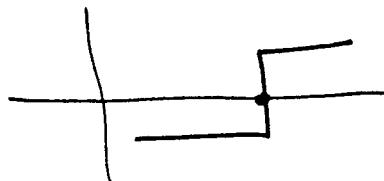
tangent:
(double root)



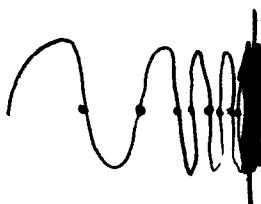
Singularity:



discontinuity:



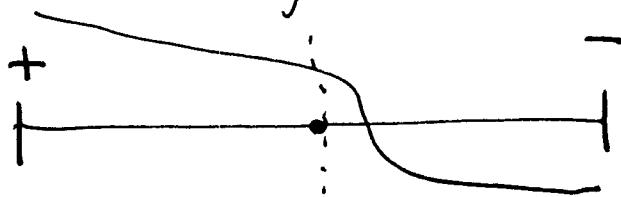
pathology:



$\sin(\frac{1}{x})$ e.g.

Bisection

Simple, linear convergence,
very reliable



bracket

evaluate at mid point
halve bracket

$$\varepsilon_{n+1} = \frac{1}{2} \varepsilon_n \quad \text{linear convergence}$$

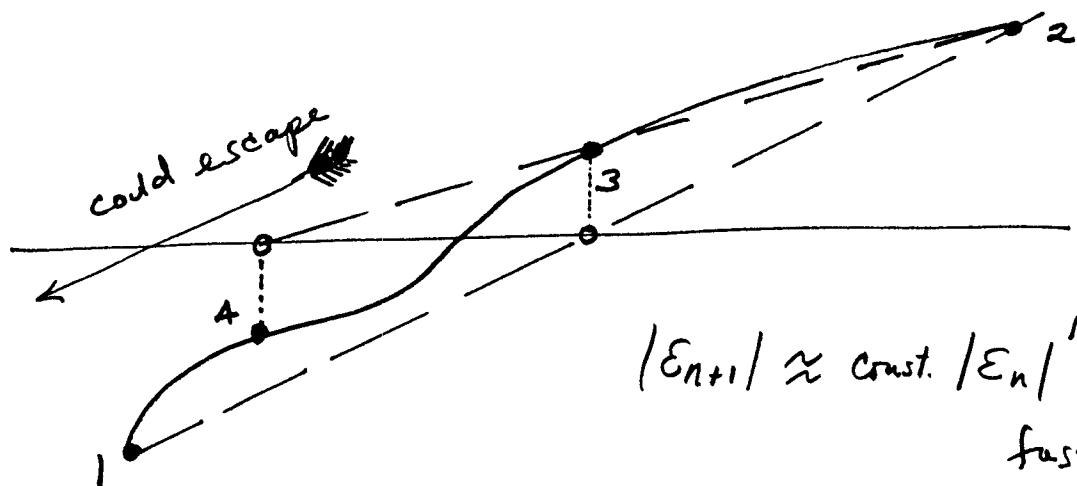
must succeed.

Tolerance for termination — requires thought, use
relative size

→ Other Classical Methods: Sometimes faster,
See [Press et al. NR] [Ralston78] [Acton70] more dangerous

Secant Method

Extrapolate or Interpolate linearly
through two most recent points.



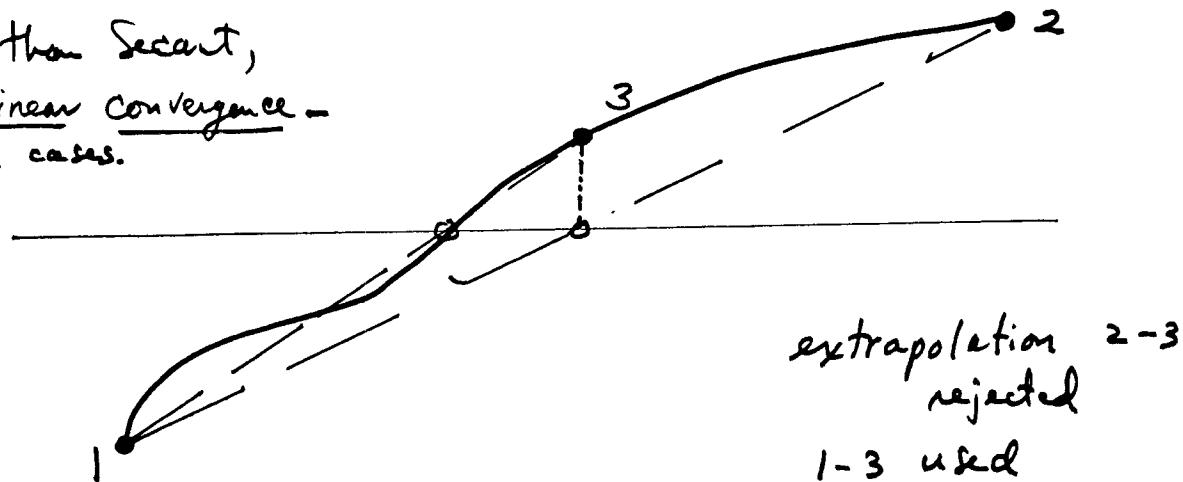
$$|\varepsilon_{n+1}| \approx \text{const. } |\varepsilon_n|^{1.618}$$

faster, but
more dangerous

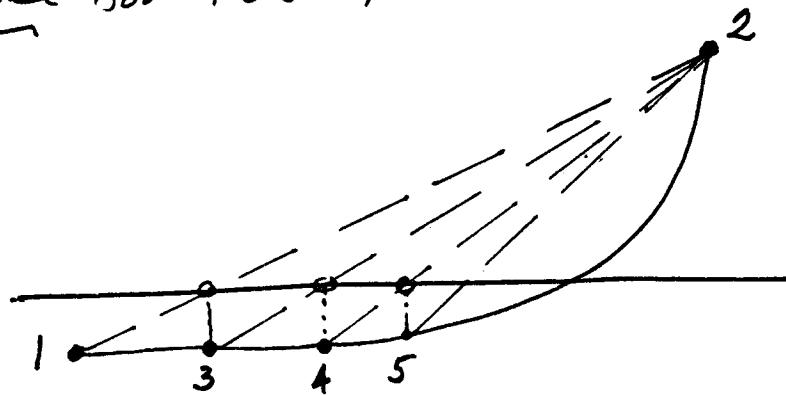
False Position (Regula falsi)

Interpolate between most recent points
that bracket root.

Safer than Secant,
 but linear convergence -
 in bad cases.



Bad Case for False Position:

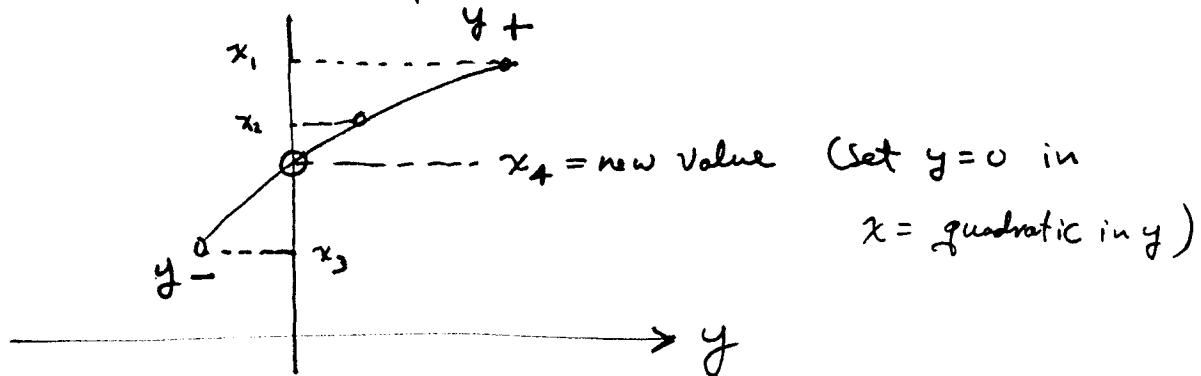


[Press et al.] Numerical Recipes recommends state-of-the-art 2.2.6

Brent's Method [Brent 73 Algorithms for Minimization Without Derivatives
Prentice-Hall]

- Combines
- root bracketing
 - bisection (if interpolation outside current bracket)
 - inverse quadratic interpolation

Inverse quadratic interpolation



Claim is superlinear (at least in smooth cases)
but safe (because of bisection backup)

All these methods avoid using derivatives.

back to Sir Isaac...

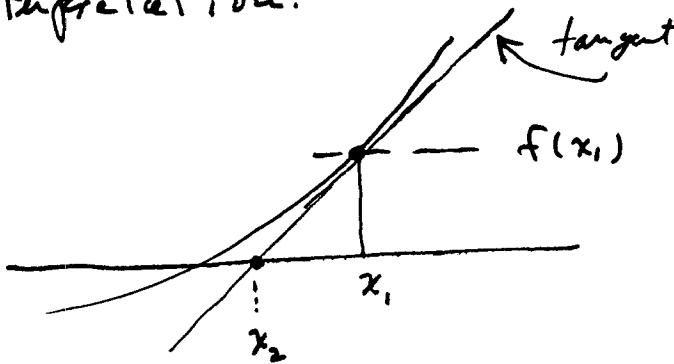
Newton-Raphson Method:

Taylor series

$$f(x+\delta) \approx f(x) + f'(x)\delta + \dots = 0$$

use $\boxed{\delta = -\frac{f(x)}{f'(x)}}$

Geometric interpretation:



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

Can be extended to n -dimensions, and is very useful — when close.

BTW, back to square-root algorithm:

$$\text{Solve } f(x) = x^2 - N = 0$$

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

$$x_{i+1} = x_i - \frac{x_i^2 - N}{(2x_i)} = \frac{1}{2} \left(x_i + \frac{N}{x_i} \right)$$

which we started with today.

In the ^{bad} old days, division was done in software

$$\frac{a}{b} = a * \left(\frac{1}{b}\right) \quad \text{So we need inverse}$$

Solve $f(x) = b - \frac{1}{x} = 0$

$$f'(x) = \frac{1}{x^2}$$

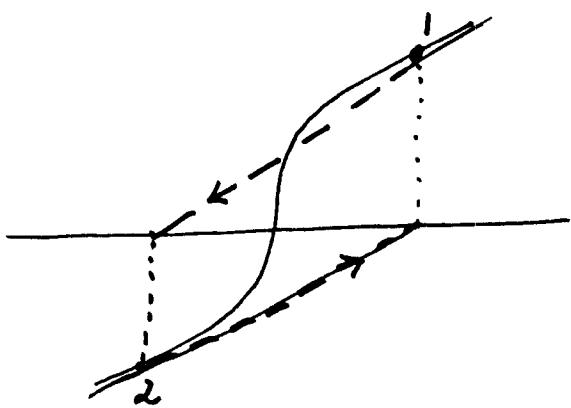
iteration $x_{i+1} = x_i - \frac{b - \frac{1}{x_i}}{\frac{1}{x_i^2}} = x_i * (2 - b * x_i)$

uses multiplication & subtraction.

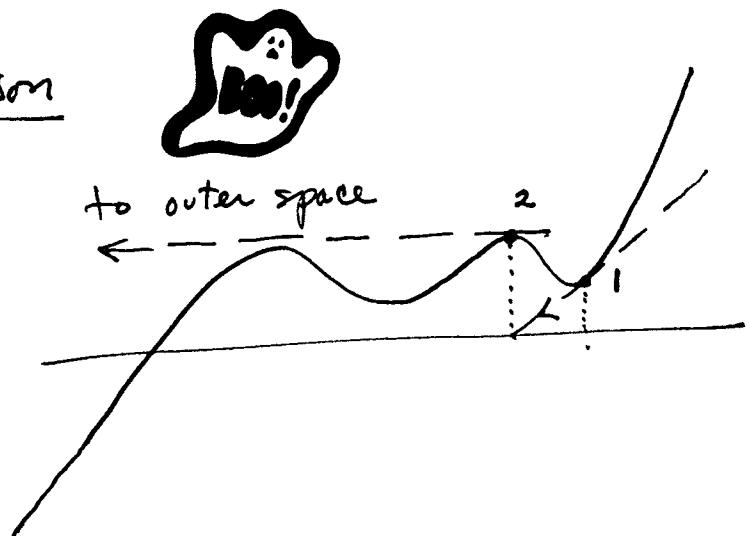
requires initial estimate $0 < x_0 < 2/b$

[Atkinson 85]

Bad Cases for Newton-Raphson



infinite loop



worse

Proof of Quadratic Convergence:

[Press et al., Atkinson, etc.]

$$f(\alpha) = f(x_n) + (\alpha - x_n) f'(x_n) + \frac{1}{2} (\alpha - x_n)^2 f''(c_n)$$

where $c_n \in [\alpha, x_n]$

and α is the desired root, $f(\alpha) = 0$

divide by $f'(x_n)$ (which can't be zero!)

$$0 = \underbrace{\frac{f(x_n)}{f'(x_n)}}_{x_n - x_{n+1}} + (\alpha - x_n) + (\alpha - x_n)^2 \frac{f''(c_n)}{2f'(x_n)}$$

$$0 = \cancel{x_n - x_{n+1}} + \cancel{\alpha - x_n} + (\alpha - x_n)^2 \frac{f''(c_n)}{2f'(x_n)}$$

$$\Rightarrow (\alpha - x_{n+1}) = \left[-\frac{f''(c_n)}{2f'(x_n)} \right] (\alpha - x_n)^2$$

$$\epsilon_{n+1} \approx \text{const. } \epsilon_n^2$$

Good Starting Point all important

Common Strategy • Start with bisection
• polish off with Newton-Raphson

Polynomial Root finding is a specialized art -
try to use highly evolved package, test answers

Zero-Finding:

2.2.10

In higher dimensions, we are in a jungle.

Insight, estimates, physical arguments - all help.

Newton-Raphson is the only general tool that's practical, but must be nursed to work on problems that are at all difficult.

There's a "Basin of Convergence":

Suppose, e.g., we want to use N-R to solve cubic

$$z^3 - 1 = 0$$

$$z_{i+1} = z_i - \frac{(z_i^3 - 1)}{3z_i^2}$$

in complex z -plane.

Mark all starting points that converge to $z=1$.

... Julia sets, fantastic pictures ...

Next topic: Optimization



Easier in general, especially in higher dimensions.