

Root finding

given some function $\vec{f}(\vec{x})$, find \vec{x}^*

$$\text{such that } \underline{\vec{f}(\vec{x}^*) = 0}$$

generally, root finders are iterative: we start from an initial guess and then successively improve it

→ do not know in advance how much work is necessary

examples of root-finding problems

magnetization of a ferromagnet

$$m = \tanh(\alpha m), \quad \alpha \text{ parameter}$$

zeta function

$$\zeta(s) = \sum_{n=1}^{\infty} \frac{1}{n^s} \quad (\text{analytically continued in the complex plane})$$

Collect \$1 mil if all non-integrals have $\operatorname{Re}(s) = \frac{1}{2}$

Eigenvalue problem

$$A\vec{x} = \lambda\vec{x}$$

both \vec{x}, λ unknown \rightarrow non-linear problem for $\begin{pmatrix} \vec{x} \\ \lambda \end{pmatrix}$

linear eigenvalue problem. Typically solved using specialized methods)

Sometimes $A(\lambda)\vec{x} = \lambda\vec{x}$

structural mechanics: $(A_2\lambda^2 + A_1\lambda + A_0)\vec{x} = 0$, \vec{x} : eigensmode
> iterative method

nonlinear BVP

$$\frac{d^2x}{dt^2} = -\frac{GM}{x^2}, \quad x(t_a) = x(t_b) = 0 \quad (\text{1D gravity})$$

\downarrow

finite difference matrix A

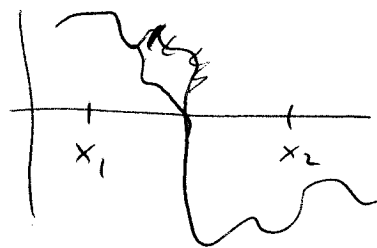
$$\rightarrow A\vec{x} = \vec{f}(\vec{x}), \quad \vec{f}(\vec{x}) = -GM \begin{pmatrix} \frac{1}{x_1} \\ \frac{1}{x_2} \\ \vdots \\ \frac{1}{x_N} \end{pmatrix} \quad \text{depends on } x_i!$$

Root finding in 1D

Bisection search

Bracket a zero: pick x_1, x_2 such that
 $\text{sign } f(x_1) \neq \text{sign } f(x_2)$

→ $f(x)$ must have a zero in between



halve the interval and compute $f(\frac{x_1 + x_2}{2})$

if $\text{sign}(f(\frac{x_1 + x_2}{2})) \neq \text{sign}(f(x_1))$, repeat the algorithm on $[x_1, \frac{x_1 + x_2}{2}]$,

otherwise repeat on $[\frac{x_1 + x_2}{2}, x_2]$

each step shrinks the interval containing the zero
by a factor of 2.

Error analysis / (convergence rate)

- saw: # of correct digits linear in
of iterations

subinterval in which the root is halved at
each iteration

Suppose initial interval is $\Delta_0 = [a, b]$

then $\Delta_i = \frac{1}{2} \Delta_0$ always

So:

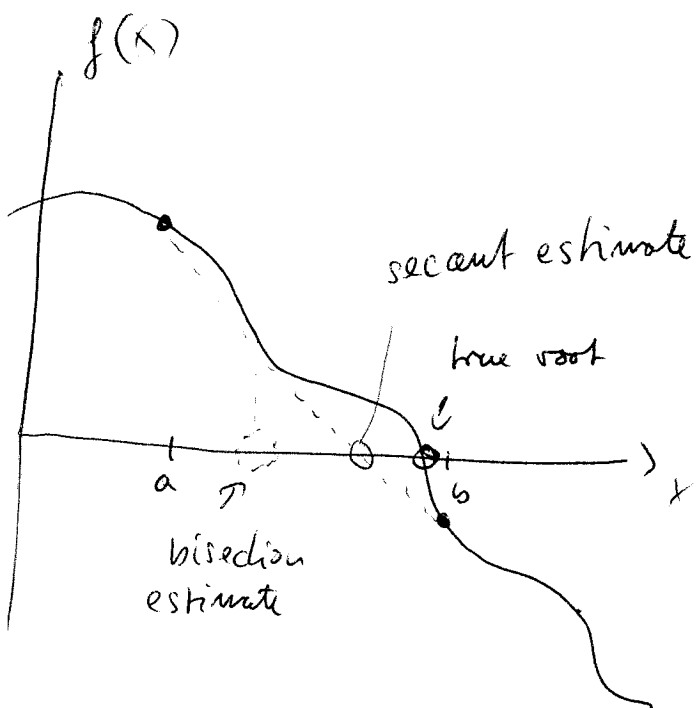
$$\Delta_n = \frac{1}{2} \Delta_{n-1} = \frac{1}{2} \cdot \frac{1}{2} \Delta_{n-2} = \dots = \frac{1}{2^n} \Delta_0$$

convergence exponentially fast (faster than anything we have
seen, N-C, finite-diff, ODE were all algebraic $\sim N^P$)

or

terminology: bisection has linear convergence (because $(\frac{1}{2})^n$
has exponent linear in n .
(compare to first-order $\sim \frac{1}{N}$)

Secant method



the bisection, bracket root on $[a, b]$, $\text{sign}(f(a)) \neq \text{sign}(f(b))$

estimate root as $x^* = b - \frac{b-a}{f(b)-f(a)} f(b)$

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

can jump out of bracket \leadsto less robust than bisection

convergence like $\sim \epsilon^p$, $p = \frac{1+\sqrt{5}}{2} \sim 1.6$

$$\epsilon_{n+1} = O(\epsilon_n^p)$$

Newton-Raphson (or often just Newton's method)

Consider secant step

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

f diff'able

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

second derivation

$$f(x) = f(x_n) + (x - x_n) f'(x_n) + O((x - x_n)^2)$$

+ $f(x^*) = 0$ we find

$$0 = f(x_n) + (x^* - x_n) f'(x_n)$$

$$\Rightarrow x^* = x_n - \frac{f(x_n)}{f'(x_n)}$$

> approximate $f(x)$ by a linear function,
then jump to that function's zero.

Convergence analysis

Let x_0 be the true root $f(x_0) = 0$

Then close to the root,

$$f(x_0) = 0 = f(x_n) + f'(x_n)(x_0 - x_n) + \frac{1}{2} f''(x_n)(x_0 - x_n)^2 + O((x_0 - x_n)^3)$$

$$\begin{aligned} \underbrace{\left(x_0 - x_n + \frac{f(x_n)}{f'(x_n)} \right)}_{\text{error}} &= -\frac{1}{2} \frac{f''(x_n)}{f'(x_n)} \underbrace{(x_0 - x_n)^2}_{= e_n^2} + O((x_0 - x_n)^3) \\ &= x_0 - x_{n+1} = e_{n+1} \end{aligned}$$

$$\Rightarrow e_{n+1} = C e_n^2 \quad (\text{quadratic convergence})$$

$$e_{n+1} \approx e_n^2 \sim (e_{n-1}^2)^2 = e_{n-1}^{2 \cdot 2} = e_{n-2}^{2 \cdot 2 \cdot 2} = \dots = e_0^{2^{n+1}}$$

number of correct digits doubles at each step

double root

double root at $x=a$

Ex: $f(x) = (x-a)^2$

→ error analysis breaks down — $f'(x_0) = 0 \rightarrow$ linear convergence

Newton's method in n-d

$$\vec{f}(\vec{x}) = \vec{0}$$

$$\vec{f}: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

↑ ↑
Newton requires here to be the same

linear case: $A\vec{x} = \vec{0}$ (special methods)

nonlinear case:

$$\vec{f}(\vec{x} + \vec{\Delta}) = \vec{f}(\vec{x}) + J\vec{\Delta} + o(\|\vec{\Delta}\|^2)$$

$$J = \begin{pmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f_m}{\partial x_1} & \dots & \frac{\partial f_m}{\partial x_n} \end{pmatrix} \quad (\text{Jacobian})$$

at a zero:

$$\vec{0} \approx \vec{f}(\vec{x}) + J\vec{\Delta}$$

$$\rightarrow \vec{\Delta} = -J^{-1}\vec{f}(\vec{x}) \quad (\text{numerically: solve system of eqns})$$

$$\rightarrow \vec{x}_{n+1} = \vec{x}_n - J^{-1}(\vec{x}_n) \tilde{f}(\vec{x}_n)$$