

## Course Overview

- handouts, syllabus, psets, etc.
  - Canvas + Piazza + Gradescope
  - [github.com/mitmath/18335](https://github.com/mitmath/18335)
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Numerical analysis:

Fast algorithms for approximately solving } math + time  
the problem of "continuous math" } + accuracy

Example: nonlinear equations

$f: \mathbb{R} \rightarrow \mathbb{R}$ ,  $f$  is continuous

Find  $f(x) = 0$

- $f$  is a general polynomial of degree  $\geq 5$ :

no root formula using  $+$ ,  $-$ ,  $\times$ ,  $\div$ ,  $\sqrt{\quad}$

- Goal: approximately find the root using only  $+$ ,  $-$ ,  $\times$ ,  $\div$

- Naive method: Bisection method

Intermediate value theorem:

$$f(a)f(b) < 0 \Rightarrow \exists \text{ root } x^* \in [a, b]$$

$$a_0 = a, \quad b_0 = b$$

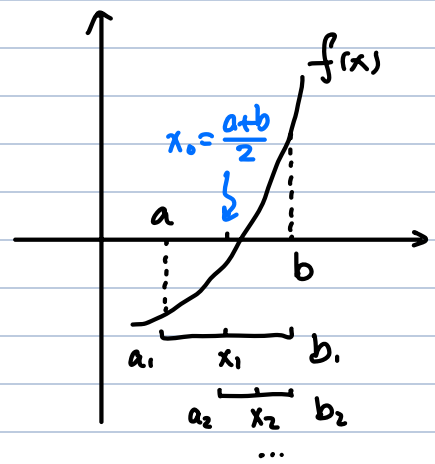
For  $n = 0, 1, 2, \dots$

$$x_n = \frac{a_n + b_n}{2}$$

$$\text{if } \text{sign}(f(x_n)) = \text{sign}(f(a_n))$$

$$a_{n+1} = x_n, \quad b_{n+1} = b_n$$

$$\text{else } a_{n+1} = a_n, \quad b_{n+1} = x_n$$



Error analysis:

$$e_n := |x_n - x^*| \leq \frac{1}{2}(b_n - a_n) = \frac{b-a}{2^n} \rightarrow 0 \quad \text{as } n \rightarrow +\infty$$

$$e_{n+1} \approx \frac{1}{2} e_n \quad \text{Linear convergence}$$

Can we do it faster? (Have only access to  $f(x)$ )

Yes. One possible method:

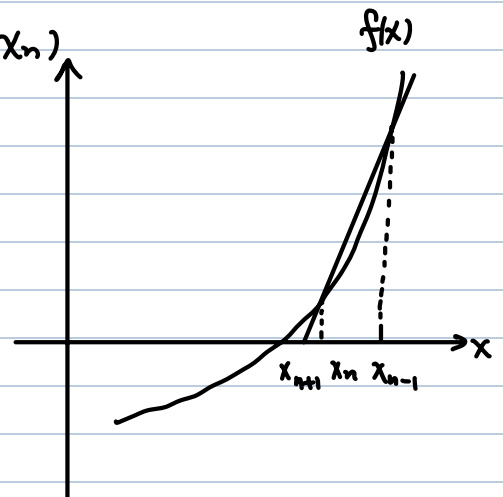
• Secant method

$$f(x) \approx f(x_n) + f'(x_n)(x - x_n)$$

$$\approx f(x_n) + \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}} (x - x_n)$$

$$= 0$$

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



Error analysis:

$$e_{n+1} := x_{n+1} - x^* = (x_n - x^*) - \frac{f(x_n)[(x_n - x^*) - (x_{n-1} - x^*)]}{f(x_n) - f(x_{n-1})}$$

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

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

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

claim:  $C_n \approx C$

$$\text{check: } f(x_n) \approx f'(x^*)e_n + \frac{1}{2}f''(x^*)e_n^2$$

$$f(x_{n-1}) \approx f'(x^*)e_{n-1} + \frac{1}{2}f''(x^*)e_{n-1}^2$$

$$\approx \frac{1}{2}f''(x^*) \underbrace{\frac{e_n - e_{n-1}}{x_n - x_{n-1}}}_{=1}$$

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

$$\approx \frac{1}{f'(x^*)} \cdot \frac{1}{2}f''(x^*) =: C$$

We arrive at

$$e_{n+1} = C e_n e_{n-1}$$

$$\text{Let } e_{n+1} \approx A e_n^p \quad (p > 1)$$

$$A e_n^p \approx C e_n \left( \frac{e_n}{A} \right)^{\frac{1}{p}} = \frac{C}{A^{\frac{1}{p}}} e_n^{1+\frac{1}{p}}$$

$$\Rightarrow A^{1+\frac{1}{p}} = C e_n^{1+\frac{1}{p}-p}$$

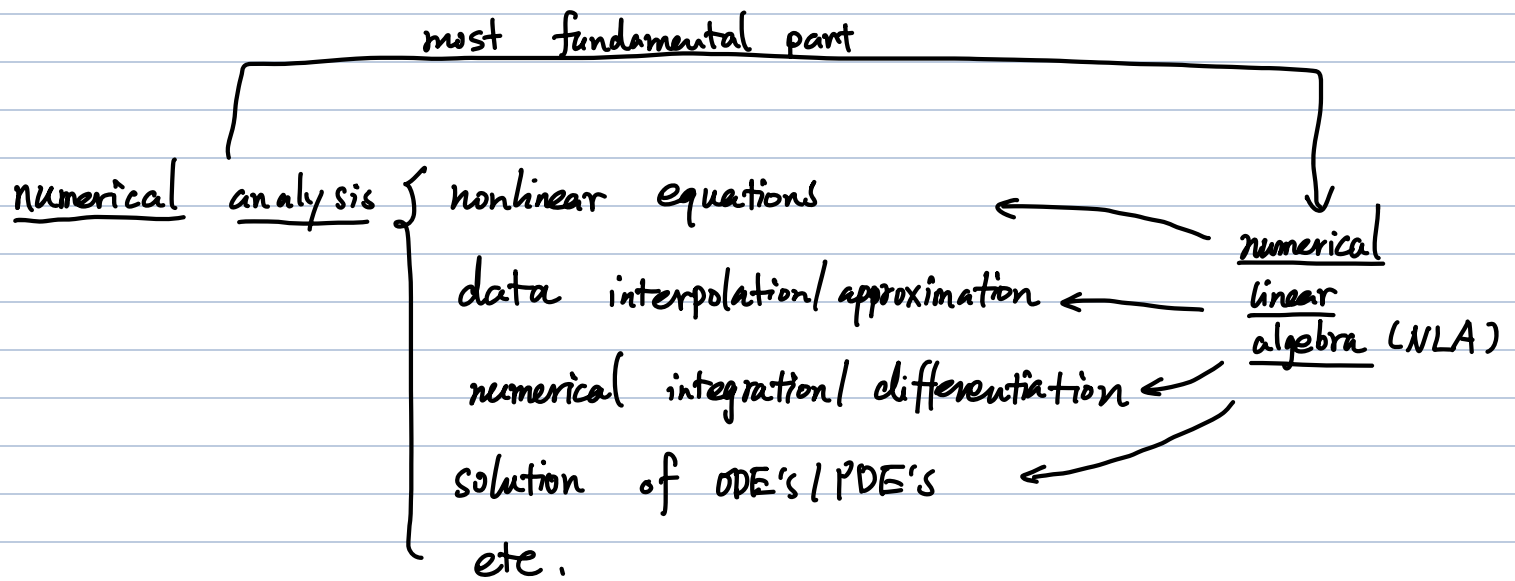
Since  $e_n \rightarrow 0$  as  $n \rightarrow +\infty$ ,

$$\text{we should have } 1 + \frac{1}{p} - p = 0 \Rightarrow p = \frac{1+\sqrt{5}}{2} \approx 1.618$$

That is  $e_{n+1} \approx C^{0.618} e_n^{1.618}$  faster than bisection

Superlinear convergence





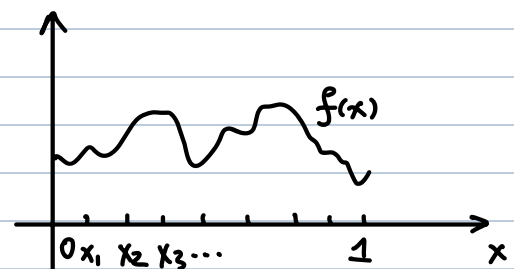
Example: Numerical methods for PDEs

a) Poisson equations

$$\begin{cases} \Delta u := \frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f \\ u|_{\partial\Omega} = 0 \end{cases}$$

In 1D,  $\frac{d^2 u}{dx^2} = f$

$u(0) = u(1) = 0$



Discretization:

$$u(x_i) \approx u_i$$

$$f(x_i) \approx f_i$$

$$h = x_{i+1} - x_i$$

$$\frac{d^2 u}{dx^2}(x_i) \approx \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2}$$

$$\Rightarrow \begin{bmatrix} -2 & 1 & & \\ 1 & -2 & 1 & \\ & & \ddots & \\ & & & 1 \\ & & & 1 & -2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} = h^2 \begin{bmatrix} f_1 \\ f_2 \\ \vdots \\ f_n \end{bmatrix}$$

Solve large linear system

$$Au = f$$

## b) Schrödinger Equation

$$-\Delta u + Vu = Eu$$

$u$ : wave function

$V$ : potential

$E$ : Quantized energy  
(eigenvalues)

• Discretization: (1D)

$$\frac{1}{h^2} \begin{bmatrix} 2+V_1h^2 & -1 & & & \\ -1 & 2+V_2h^2 & & & \\ & & \ddots & & \\ & & & -1 & \\ -1 & & & 2+V_nh^2 \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix} = E \begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{bmatrix}$$

$$Au = Eu$$

Solve large eigenvalue problems

NLA

### Outline

fundamentals of  
numerical analysis

Floating-point arithmetic  
matrix norms  
backward error analysis  
matrix factorization, etc.

Direct method: Gauss/Cholesky  
Iterative method: Jacobi/G-S.  
etc.

Given  $A \in \mathbb{R}^{n \times m}$ ,  
 $b \in \mathbb{R}^n$

Solve  $Ax = b$

Given  $A \in \mathbb{R}^{n \times n}$

Find all  $(\lambda, x)$   
s.t.  $Ax = \lambda x$

Power method  
QR iteration

Krylov subspace methods

Arnoldi  
GMRES  
CG / BiCG, etc.

Other topics: Nonlinear optimization  
Monte Carlo sampling  
Randomized algorithms etc.