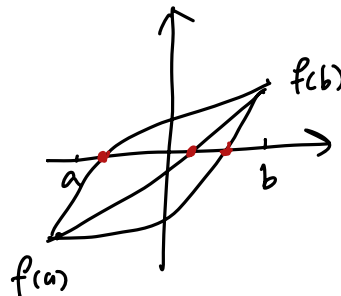


Midterm Review

1. Solution of Equations = solve for $f(x)=0$ for $x \in [a, b]$

① existence of such solution

• if $f(a) \cdot f(b) < 0$, then $\exists x \in [a, b]$ s.t. $f(x)=0$.
(assume f is cts)



② Fixed point iteration.

Solving for $f(x)=0 \Leftrightarrow g(x)=x$, for suitable choice of g .

Given $x_0 \in [a, b]$, $x_{k+1} = g(x_k)$, $k=0, 1, 2, \dots$

②.1 Convergence of f_p

(strongest) If $\exists L < 1$, s.t. $\forall x, y \in [a, b]$, $|g(x) - g(y)| \leq L \cdot |x - y|$,
then it converges regardless of starting point.

If $g \in C^1$, then this is equivalent to $\sup_{x \in [a, b]} |g'(x)| \leq L$.

(weaker) Let x be the fixed point of g .

If $g \in C^1$ and $|g'(x)| < 1$, then it converges if we start sufficiently close to x .

Ex: Let $f(x) = xe^x - 1$. Show that $f(x)$ has unique root on \mathbb{R} .

Sln: Existence of root: $f(0) = 0 - 1 = -1$

$$f(1) = 1 \cdot e^1 - 1 = e - 1 > 0.$$

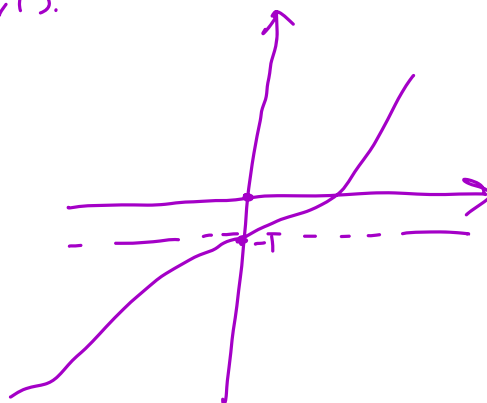
Since $f(0) \cdot f(1) < 0$, \exists a zero of f in $(0, 1)$.

Uniqueness: For $x \leq 0$, $xe^x \leq 0$

$$\Rightarrow f(x) = xe^x - 1 \leq 0 - 1 = -1$$

For $x > 0$, $f'(x) = (xe^x - 1)'$

$$= e^x + xe^x = (x+1)e^x > e^x > 0$$



$\Rightarrow f'(x) > 0 \Rightarrow f$ is always increasing on $(0, \infty)$.

$\Rightarrow f$ only crosses the x -axis once \Rightarrow soln of $f(x)=0$ is unique.

$$\varphi(x) = x - \lambda f'(x)$$

ex: $x_{k+1} = \varphi(x_k) = x_k - \lambda f'(x_k)$, $\lambda > 0$, $f \in C^2$

$$x_{k+1} = g(x_k)$$

\wedge fixed pt.

$$x = g(x)$$

① What are fixed points of φ ?

$$x = \varphi(x)$$

$$x = x - \lambda f'(x) \Rightarrow f'(x) = 0$$

② Assume $f''(x) > c > 0 \forall x$. What choice of λ will the fixed point iteration converges for any initial condition x_0 ?

soln: We need $|\varphi'(x)| < 1 \quad \forall x$

$$\varphi'(x) = 1 - \lambda f''(x) < 1 - \lambda \cdot c$$

$$\Rightarrow \text{need } \begin{cases} 1 - \overset{>0}{\lambda c} < 1 \Rightarrow \text{automatic} \\ \underbrace{1 - \lambda c > -1}_{\lambda > 0} \Rightarrow -\lambda c > -2 \Rightarrow \lambda < \frac{2}{c}. \end{cases}$$

④ Iterative method for root finding.

1° Newton's method.

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)}, \quad k=0, 1, \dots \quad f'(x_k) \neq 0$$

If it converge, and $f'(x) \neq 0$, quadratic convergence speed.

--- and $\underbrace{f'(x) = 0}$, linear convergence.

x here satisfies $f(x) = 0$.

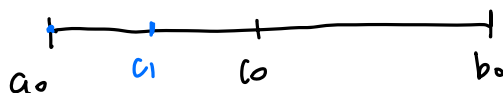
2° Secant Method.

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

3° Bisection method.

If \exists a zero in some interval, then f evaluated at the interval endpoints have negative signs.

Idea: only focusing on intervals that have different signs at the endpoints. cut the interval half again.



2. Computer representation of number.

• fixed point = 0.000244 / 12345.789

• Floating point representation:

$$\left\{ \begin{array}{l} 0.000244 = 2.44 \times 10^{-4} \\ 12345.789 = \underline{1.2345789} \times 10^4 \end{array} \right.$$

$$1 \leq x < 10$$

$$\pm m \times 2^E, \quad 1 \leq m < 2$$

↗ sign ↑ significand/
mantissa

← exponent

- Machine precision: single: $2^{-23} \approx 10^{-7}$
 double: $2^{-52} \approx 10^{-16}$

3. Solutions of systems of linear equations. Solve the system $Ax=b$

① Existence of the solution:
and unique if A is invertible.

② LU decomposition.

$$A=LU, \quad L = \begin{pmatrix} 1 & & 0 \\ x & 1 & \\ \vdots & \ddots & \ddots \\ x & \dots & x & 1 \end{pmatrix}, \quad U = \begin{pmatrix} x & x & \dots & x \\ & \ddots & & \vdots \\ & & & x \end{pmatrix}$$

$$\text{Given } A=LU, \quad Ax = \underbrace{LUx}_y = b \quad \left\{ \begin{array}{l} Ly = b \quad \text{forward} \\ Ux = y \quad \text{backward.} \end{array} \right.$$

Cost of getting the LU: $\frac{2}{3}n^3 - \frac{1}{2}n^2 = O(n^3)$

--- solving forward/backward: $2n^2 + \dots = O(n^2)$

Eg: Manually calculate LU.

$$A = \begin{pmatrix} 4 & 2 & 0 & 0 \\ 2 & 5 & 4 & 0 \\ 0 & 4 & 6 & 3 \\ 0 & 0 & 3 & 6 \end{pmatrix}$$

$$L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 1/2 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 \\ 0 & 0 & 3/2 & 1 \end{pmatrix}$$

$$\frac{a_{21}}{a_{11}} = \frac{2}{4} = \frac{1}{2}$$

$$\begin{pmatrix} 4 & 2 & 0 & 0 \\ 0 & 4 & 4 & 0 \\ 0 & 4 & 6 & 3 \\ 0 & 0 & 3 & 6 \end{pmatrix}$$

$$\frac{a_{32}}{a_{22}} = \frac{4}{4} = 1$$

$$\begin{pmatrix} 4 & 2 & 0 & 0 \\ 0 & 4 & 4 & 0 \\ 0 & 0 & 2 & 3 \\ 0 & 0 & 3 & 6 \end{pmatrix}$$

$$\frac{a_{43}}{a_{33}} = \frac{3}{2}$$

$$\begin{pmatrix} 4 & 2 & 0 & 0 \\ 0 & 4 & 4 & 0 \\ 0 & 0 & 2 & 3 \\ 0 & 0 & 0 & 3/2 \end{pmatrix} = U$$

(3) Pivoting: $PA = LU \Leftrightarrow A = P^T LU$ b/c $P^T = P^{-1}$

$$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix} \begin{matrix} \updownarrow \\ \updownarrow \end{matrix} \text{ swap the rows.}$$

$$P = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{matrix} \updownarrow \\ \updownarrow \end{matrix} \Rightarrow P = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$$

4. Norm and Conditioning.

Given $Ax=b$, how sensitive is x to changes in A/b ?

① Conditioning: Condition numbers of matrix A .

↳ small changes in b could cause large changes in x .

$$A(x+\delta x) = b + \delta b.$$

$$\Rightarrow \frac{\|\delta x\|}{\|x\|} \leq K(A) \cdot \frac{\|\delta b\|}{\|b\|}$$

↑
condition number of A .

$K(A) = \|A\| \cdot \|A^{-1}\|$, the choice of norm is 1-norm, 2-norm, ∞ -norm.

② Is this a norm?

Is $|x|$ a norm? for $x \in \mathbb{R}^2$, $x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$

No. For norm, $\|x\| \geq 0$, and $\|x\|=0$ iff $x=0 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$

Let $x = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$, then $|x|=0$, but $x \neq \begin{pmatrix} 0 \\ 0 \end{pmatrix}$.

5 Least Squares

Solve the over-determined system $Ax=b$

$$\begin{matrix} m & n \\ (m > n) & \end{matrix} \begin{matrix} \boxed{A} \\ \end{matrix} \begin{matrix} 1 \\ n \\ \boxed{x} \end{matrix} = \begin{matrix} 1 \\ m \\ \boxed{b} \end{matrix}$$

① Least square formulation: $\min_x \|Ax-b\|_2$

$$\Leftrightarrow \min_x \|Ax-b\|_2^2$$

$$\Leftrightarrow \min_x \frac{1}{2} \|Ax-b\|_2^2$$

② Normal equations.

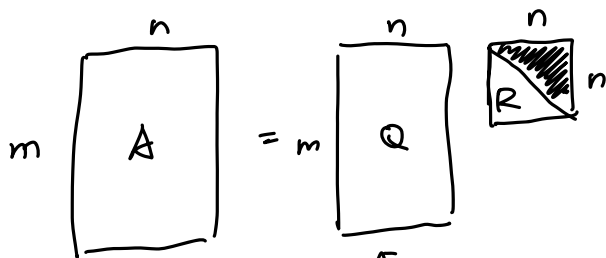
Sln of least square method solves the normal equation

$$A^T A x = A^T b$$

→ if A is full rank, i.e., $\text{rank}(A) = n$, then $A^T A$ is invertible,
and \exists unique sln x .

6. QR factorization

\wedge done on rectangular / square matrix.



Q has orthonormal columns.

If A is square, then Q is also square, Q is orthogonal,

$$Q^T Q = Q Q^T = I$$

$$\Rightarrow Q^T = Q^{-1}$$

• obtaining QR: Gram Schmidt on columns of A .