

T63.1050

Prof. Papanicolaou

ELEM. NUMERICAL METHODS

Books of Interest

- ALGORITHM 1. Conte-de Boor - Elementary Numerical Anal. Course Text. 1972
 2. P. Henrichi, Elements of Numerical Anal 1964
 RIGOROUS * 3. Isaacson & Keller - Analysis of Numerical Methods 1966

Recitation Sessions: Tues 4-5 Wed 2-3 STAMPE { 4/1 TECH 1

LAST WK BEFORE EXAMS IS READING WEEK.

Numbers & Bases.

$$(53781)_{10} = 5 \cdot 10^4 + 3 \cdot 10^3 + 7 \cdot 10^2 + 8 \cdot 10 + 1 \cdot 10^0$$

$$(a_n a_{n-1} \dots a_0)_{10} = a_n \cdot 10^n + a_{n-1} \cdot 10^{n-1} + \dots + a_1 \cdot 10^1 + a_0 \cdot 10^0 \quad 0 \leq a_j \leq 9$$

Binary repres. of integers

$$N = (a_n \dots a_0) = a_n \cdot 2^n + a_{n-1} \cdot 2^{n-1} + \dots + a_1 \cdot 2^1 + a_0 \cdot 2^0 \quad a_j = 0, 1$$

$$(1001)_2 = (9)_{10}$$

Algorithm of Polynomial Evaluation by nested operation

$$\text{Let } p(x) = \sum_{i=0}^n a_i x^i$$

Evaluate $p(x)$ at $x = \beta$

$$p(x) = (\dots((a_n x + a_{n-1})x + a_{n-2})x + a_{n-3}) \dots)x + a_0$$

To evaluate $p(\beta)$ evaluate $0 \leq a_j \leq 9 \quad b_n \dots b_i \dots b_0$

$$b_n = a_n$$

$$b_{n-1} = b_n \beta + a_{n-1}$$

$$b_{n-2} = b_{n-1} \beta + a_{n-2}$$

$$\vdots$$

$$b_0 = b_1 \beta + a_0$$

$$b_0 = p(\beta)$$

$$(11100101)_2 = ? \text{ decimal}$$

Apply Algo $n=7$

$$b_7 = 1$$

$$b_6 = 1 \cdot 2 + 1 = 3$$

$$b_5 = 3 \cdot 2 + 1 = 7$$

$$b_4 = 7 \cdot 2 + 0 = 14$$

$$b_3 = 14 \cdot 2 = 28$$

$$b_2 = 28 \cdot 2 + 1 = 57$$

$$b = 57.2 = 114$$

$$b_0 = 114 \cdot 2 + 1 = 229$$

$$(11100101)_2 = (229)_{10}$$

Convert decimal to binary

$$(229)_{10} = ?_2 \quad 2 \cdot 10^2 + 2 \cdot 10 + 9 \quad (10)_{10} = (1010)_2$$

$$(10)_2 (1010)_2^2 + (10)_2 (1010)_2 + (1001)_2$$

$$(1010)_2^2 = (1100100)_2$$

$$(10)_2 (1010)_2^2 = \begin{array}{r} 11001000 \\ 10100 \\ \hline 1001 \\ 11100101 \end{array}$$

Arbitrary Real Numbers

$$x = (a_n a_{n-1} \dots a_1 a_0 . b_1 b_2 b_3 \dots)_2 \quad 0 \leq a_j \leq 9$$

$$x = a_n \cdot 10^n + a_{n-1} \cdot 10^{n-1} + a_{n-2} \cdot 10^{n-2} + \dots + a_0 + b_1 \cdot 10^{-1} + b_2 \cdot 10^{-2} + \dots + b_n \cdot 10^{-n}$$

$$X_{\text{INT}} = (a_n a_{n-1} \dots a_0)$$

$$X_{\text{FRAC}} = (b_1 b_2 \dots)$$

$$x = X_{\text{I}} + X_{\text{F}}$$

Conversion of Decimal Fractions to Binary Fractions

$$x_F = a_1/10 + a_2/10^2 + \dots + a_n/10^n$$

$$b_1/2 + b_2/2^2 + \dots + b_n/2^n$$

$$2x_F = b_1 + b_2/2 + b_3/2^2 + \dots + b_n/2^{n-1}$$

$$(2x_F)_F = b_2/2 + b_3/2^2 + \dots \quad \text{Integ part of } 2x_F = b_1$$

$$= (2x_F - b_1)$$

$$b_2 = \text{Integral part} = 2(2x_F - b_1)$$

.65321 $\rightarrow$? binary

$$2(.65321) = 1.30642$$

$$b_1 = 1$$

$$2(1.30642) = 2.61284$$

$$b_2 = 0$$

$$b_3 \geq 1$$

$$2(6128) = 12256$$

$$(65321)_{10} = (.101\dots)_2$$

Floating Point Representation of Real Number

$$x \rightarrow fp(x)$$

x = floating point in exact decimal expansion

$$fp(x) = \pm (d_1 d_2 \dots d_n) \cdot b^e \quad \text{for base } b, \quad 0 \leq d_i \leq b-1$$

mantissa

exponent

$$m \leq e \leq M \quad m = -M$$

n = no. of significant figures

$$cde \ 6600 \quad n=38 \quad b=2 \quad \pm 10^{\pm 400} \quad d_1 \neq 0$$

$$|x| > \beta^M \text{ overflow}$$

$$0 < |x| < \beta^{m+1}$$

Given: $fp(x+y) \neq fp(x) + fp(y)$ na.

$$\left\{ \begin{array}{l} \text{rounding} \\ \text{dropping} \end{array} \right. \quad \begin{array}{l} \text{dn does not change} \\ \text{dn} \rightarrow \text{dn}+1 \end{array} \quad \begin{array}{l} \text{dn}+1 \geq \beta/2 \\ \text{dn}+1 \geq \beta/2 \end{array}$$

$$\text{Round off error} = |x - fp(x)| = \frac{1}{2} \beta^{1-n}$$

$$x = .666\dots \quad n=3 \quad \beta=10$$

$$fp(x) = .667 \times 10^0 \quad \text{Round off} = .666\dots = .66700 < .5 \times 10^{-2}$$

Warning: loss of significant figures

$$\begin{array}{l} x_{\#} = fp(x) = .123456789 \quad \# \text{ sig figures} = 9 \\ y_{\#} = fp(y) = .123456781 \quad \# \text{ sig figures} = 9 \\ z_{\#} = x_{\#} - y_{\#} = .8 \times 10^{-8} \quad \# \text{ of sig figures} = 1 \end{array}$$

$$\text{Ex: } ax^2 + bx + c = 0 \quad a = 6 \times 10^8, \quad b = -6.7 \times 10^8, \quad c = 24 \quad b^2 \gg 4ac$$

for stability

$$ax^2 + bx + c = a(x^2 + \frac{b}{a}x + \frac{c}{a})$$

$$= a(x - x_1)(x - x_2)$$

$$= a[x^2 - (x_1 + x_2)x + x_1x_2]$$

$$x_1 + x_2 = -\frac{b}{a}$$

$$x_1x_2 = \frac{c}{a}$$

$$x_{1,2} = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

Order Symbols

if $a_1, a_2, a_3, \dots$ is a seq of real numbers
 then $a_n \rightarrow a$ if for any $\epsilon > 0$ $|a_n - a| < \epsilon$
 provided $n > n_0(\epsilon)$

Speed of convergence

If $b_1, b_2, \dots$ is another seq suppose $b_n \rightarrow 0$ as $n \rightarrow \infty$
 examples: $b_n = 1/n$ or $1/n^2$ where $0 < \delta < 1$
 Comparison Sequence

we write

$$a_n = a + o(b_n)$$

if $\left| \frac{b_n}{a_n - a} \right| \leq k$ some constant $\forall$ large n .

Example:

$$\text{Let } a_n = \sum_{k=0}^n r^k \quad |r| < 1 \quad a_n \rightarrow \sum_{k=0}^{\infty} r^k = \frac{1}{1-r} = a$$

$$a_n = \frac{1-r^{n+1}}{1-r} = a + \left(\frac{r^{n+1}}{1-r} \right) = a + o(r^n) \quad |r| < 1$$

if the above is true then $a_n \rightarrow a$ geometrically fast.

Definition

if $\frac{|a_n - a|}{b_n} < \epsilon$ where $\epsilon > 0$ for n sufficient large n

then $a_n = a + o(b_n)$ goes to a faster than b_n

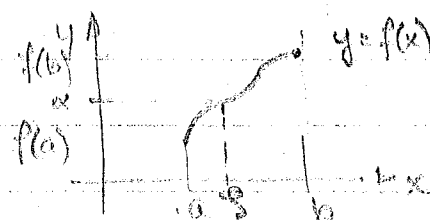
$$a_n = a + o(b_n)$$

$$c_n = c + o(b_n)$$

$$a_n + c_n = a + c + o(b_n)$$

$$a_n c_n = ac + o(b_n) \quad \text{if } a^2 + c^2 \neq 0 \quad \text{if } a+c \text{ are } 0 \text{ then } o(b_n^2)$$

Intermediate Value Theorem



if $f(a) < \alpha < f(b)$ & f is continuous
then $\exists a \leq s \leq b$ s.t. $f(s) = \alpha$

Taylor's Theorem

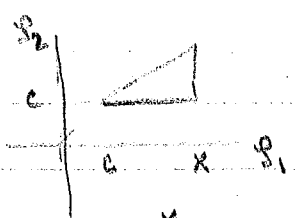
let $f(x) \in C^{n+1}$ in $[a, b]$

$$\text{then } f(x) = f(c) + \sum_{k=1}^n \frac{f^{(k)}(c)}{k!} (x-c)^k + R_{N+1}$$

$$R_{N+1} = \frac{1}{n!} \int_c^x f^{(n+1)}(x) (x-c)^n dx$$

Note that $f(x) = f(c) + \int_c^x f'(s_1) ds_1$ Fundamental theorem of Calculus
 $f'(s_1) = f'(c) + \int_c^{s_1} f''(s_2) ds_2$

$$f(x) = f(c) + \int_c^x \left[f'(c) + \int_c^{s_1} f''(s_2) ds_2 \right] ds_1 = f(c) + f'(c) [x-c] + \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1$$



$$\int_c^x \int_{s_2}^x f''(s_2) ds_1 ds_2 = \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1$$

$$\int_c^x (x-s_2) f''(s_2) ds_2$$

HW

Verify by integration

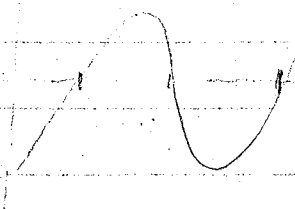
$$\int_c^x \int_c^{s_1} \int_c^{s_2} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n \dots ds_1$$

Prob 1-32, 33, 43, 41

1-76

locate by computer: roots of $f(x) = x^3 - 6x^2 + 11x - 6$ $[1, 2, 3]$

by dissection, ordinary regula falsi & newton



problem 2.14, 11 pg 34

2 Oct. 1972

1. Review $fp(\cdot)$ floating point

2. Solving NON-LINEAR EQS.

x a real number

$$x = \pm (.d_1 d_2 \dots) \beta^e \quad 0 \leq d_1 \leq \beta-1 \quad d_1 \neq 0$$

$$fp(x) = \pm (. \delta_1 \delta_2 \dots \delta_n) \beta^e \quad \text{Two methods of floating a number}$$

1. Chopping $\delta_1 \delta_2 \dots \delta_n = d_1 d_2 \dots d_n$

2. Rounding

$$(i) \quad d_{n+1} < \beta/2 \quad \delta_1 \delta_2 \dots \delta_{n-1} \delta_n = d_1 \dots d_n$$

$$(ii) \quad d_{n+1} \geq \beta/2 \quad \delta_1 \delta_2 \dots \delta_{n-1} = d_1 \dots d_{n-1}; \delta_n = d_{n+1}$$

Remark

Assume $x \geq 0$

$$(i) \quad x \gg x^* = fp(x) \text{ by chopping and } \frac{|x - x^*|}{|x|} \leq \beta^{-(1-n)}$$

$$(ii) \quad \text{In rounding } x > x^* \text{ or } x \leq x^*$$

$$\frac{|x - x^*|}{|x|} \leq \frac{1}{2} \beta^{-(1-n)}$$

$$\text{Verify (i)} \quad |x - x^*| = (. \underset{n}{\underbrace{000 \dots 0}_{n}} d_{n+1} d_{n+2} \dots) \beta^e = (. d_{n+1} d_{n+2} \dots) \beta^{e-n}$$

$$= \frac{(. d_{n+1} d_{n+2} \dots)}{(. d_1 d_2 \dots)} (d_1 d_2 \dots) \beta^e \beta^{-n}$$

$$= \frac{(. d_{n+1} d_{n+2} \dots)}{(. d_1 d_2 \dots)} |x| \beta^{-n} \leq \frac{(. \beta-1, \beta-1, \dots)}{(. 1000 \dots)} |x| \beta^{-n}$$

largest no. can be

smallest no. can be

$$\leq \beta |x| \beta^{-n} = |x| \beta^{1-n}$$

Order Symbols - $\alpha_1, \alpha_2, \dots$ $\alpha_n \rightarrow \alpha$

$$\alpha_n = \alpha + O(\beta_n) \iff \left| \frac{\alpha_n - \alpha}{\beta_n} \right| \leq K \text{ for sufficiently large } n$$

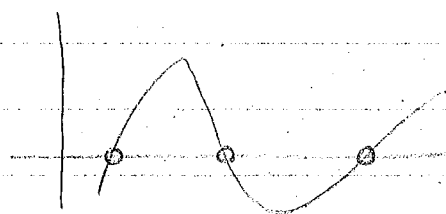
$$\alpha_n = \alpha + O(\beta_n) \iff \left| \frac{\alpha_n - \alpha}{\beta_n} \right| \leq \epsilon \text{ for every choice of } \epsilon > 0$$

provided $n = n(\epsilon)$ is sufficiently large.

$$\alpha_n = \alpha + O(\beta_n) \iff \alpha_n \sim \alpha + K\beta_n$$

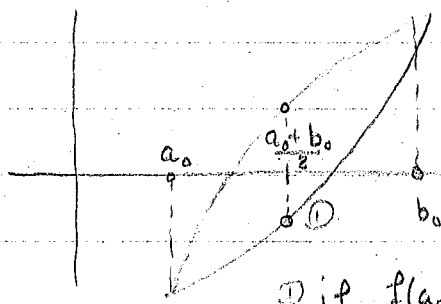
Solution of non-linear Equations

$f(x)$ is a given function.



find the roots s.t. $f(x) = 0$
 $x \in \mathbb{R}$ and $f \in \mathbb{R}$

① Method of Bisection



Given $f(x)$ s.t. $f(a_0) < 0 < f(b_0)$

begin with $(a_0, b_0) \rightarrow$ construct $M = \frac{a_0 + b_0}{2}$

check $f(a_0) f(m)$

① if $f(a_0) f(m) > 0$ then set $a_1 = m, b_1 = b_0$ repeat

② if $f(a_0) f(m) \leq 0$ set $a_0 = a_1, b_1 = m$

Suppose a_0, b_0 are points s.t. the continuous $f(x)$ satisfies $f(a_0) < 0 < f(b_0)$

then for $n = 0, 1, 2, \dots$ Do

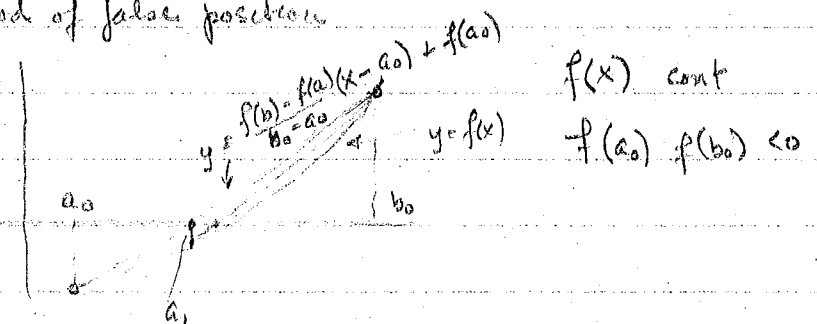
set $m = \frac{a_n + b_n}{2}$

if $f(a_0) f(m) > 0$ set $a_{n+1} = m, b_{n+1} = b_n$

≤ 0 set $a_{n+1} = a_n, b_{n+1} = m$

1. doesn't use properties of function

② method of false position



$$-f(a_0) = \frac{f(b_0) - f(a_0)}{b_0 - a_0} (a_1 - a_0)$$

$$a_1 = \frac{f(b_0)a_0 - f(a_0)b_0}{f(b_0) - f(a_0)} \quad (\text{numerically stable})$$

ALGORITHM :

If $f(a_0)f(b_0) < 0$ then for $n = 0, 1, 2, \dots$ DO

$$\text{Compute } m = \frac{f(b_n)a_n - f(a_n)b_n}{f(b_n) - f(a_n)}$$

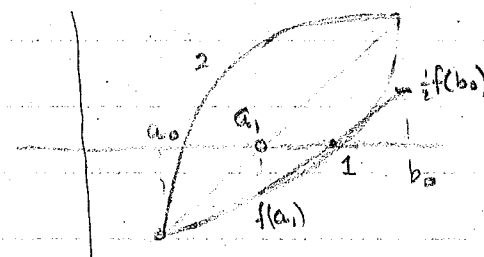
If $f(a_n)f(m) > 0$ set $a_{n+1} = m, b_{n+1} = b_n$

≤ 0 set $a_{n+1} = a_n, b_{n+1} = m$

then $[a_{n+1}, b_{n+1}]$ contains a root of $f(x)$

1. does not use the fact that one of the interval endpoints doesn't move

③ Method of Modified Regula Falsi



Statement of Algorithm:

Suppose $f(x)$ is cont.

and $f(a_0)f(b_0) < 0$, set $W_0 = a_0$ & $F = f(a_0)$ $G = f(b_0)$

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

Compute $W_{n+1} = \frac{Ga_n - Fb_n}{G - F}$

1 If $f(a_n)f(W_{n+1}) > 0$, set $a_{n+1} = W_{n+1}$, $b_{n+1} = b_n$, $F = f(W_{n+1})$

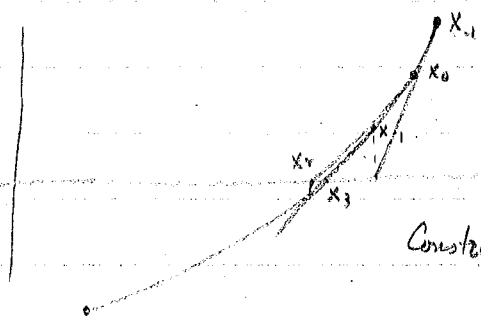
If also $f(W_n)f(W_{n+1}) > 0$, set $G = G/2$ repeat.

2 If $f(a_n)f(W_{n+1}) < 0$, set $a_{n+1} = a_n$, $b_{n+1} = W_{n+1}$ $G = f(W_{n+1})$

If also $f(W_n)f(W_{n+1}) > 0$ set $F = F/2$ repeat

Then $[a_{n+1}, b_{n+1}]$ contains a root of $f(x)$

④ Secant Method



Given x_1, x_0

$f(x)$ continuous

then for $n = 0, 1, 2, \dots$ DO

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

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

Then $\dots, x_n \rightarrow \text{root}$

⑤ Newton Method

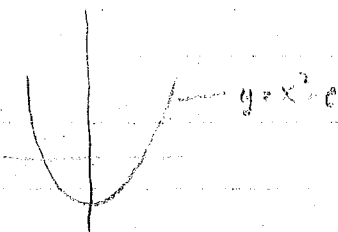
Assume f is diff & x_0 is given then for $n = 0, 1, \dots$

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

then $x_n \rightarrow \text{root}$

Example: (Newton)

Given $c > 0$ find $\sqrt{c}$ let $f(x) = x^2 - c$



$$y = x^2 - c$$

$$y' = 2x$$

$$y = x^2 - c$$

By Newton's method $x_{n+1} = x_n - \frac{(x_n^2 - c)}{2x_n}$
 $= \frac{1}{2} \left(x_n + \frac{c}{x_n} \right)$

Try to find $\sqrt{2} \approx 1.414214$

let $x_0 = 1$ $x_1 = \frac{1}{2} \left(1 + \frac{2}{1} \right) = \frac{3}{2}$

$$x_2 = \frac{1}{2} \left(\frac{3}{2} + \frac{2}{3/2} \right) = \frac{17}{12} \approx 1.41$$

$$x_3 = \frac{1}{2} \left(\frac{17}{12} + \frac{2}{17/12} \right)$$

$$\sqrt{2} \approx 1.732$$

$$x_0 = 1$$

$$x_3 = \frac{97}{56} \approx 1.732 \quad \text{Root}$$

Some theoretical Analysis

$$f(x) = 0$$

$$x \in [a, b]$$

For there purpose better to study $x = g(x)$, $[a, b]$

For example $g(x) = x - h(x)f(x)$ $h(x) > 0$ in $[a, b]$

How to find fixed points of $g(x)$ i.e. $\xi \in [a, b]$ s.t. $g(\xi) = \xi$

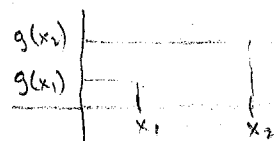
let $x_0 \in [a, b]$ & construct

$$x_{n+1} = g(x_n)$$

If $x_n \rightarrow$ convergence then $x_{n+1} \rightarrow \xi$, $g(\xi) = \xi$

Theorem:

⊗ let $g(x)$ map $[a, b] \rightarrow [a, b]$ and assume $|g(x_1) - g(x_2)| \leq K|x_1 - x_2|$
 $\forall x_1, x_2 \in [a, b]$ $0 < K < 1$ $g(x)$ is called a contraction



then $x_{n+1} = g(x_n)$ $x_0 = \text{arbitrary} \in [a, b]$

converges to a unique fixed point

$\mathcal{S}$ of $g(x)$.

Remark: if $|g'(x)| \leq K < 1$ for $\forall x$ in $[a, b]$ then $\odot$ is true.

Proof:

To show $[x_n]$ is a Cauchy seq i.e. given $\epsilon > 0$ $|x_m - x_n| < \epsilon$

$\forall$ suff large $m \geq n(\epsilon)$

From a theorem in calc then x_n converges to a point $\mathcal{S}$ in $[a, b]$

$\mathcal{S}$ is unique.

claim: $|x_m - x_n| \leq \frac{K^n}{1-K} |x_1 - x_0| \quad \forall m$

$$|x_m - x_n| = |(x_m - x_{m-1}) + (x_{m-1} - x_{m-2}) + \dots + (x_{n+1} - x_n)|$$

$$\leq |x_m - x_{m-1}| + |x_{m-1} - x_{m-2}| + \dots + |x_{n+1} - x_n|$$

$$|x_{j+1} - x_j| = |g(x_j) - g(x_{j-1})| \leq K |x_j - x_{j-1}| = K |g(x_{j-1}) - g(x_{j-2})|$$
$$\leq K^2 |x_{j-1} - x_{j-2}|$$

$$\vdots$$
$$\leq K^j |x_1 - x_0| \quad \text{geometric}$$

$$|x_m - x_n| < (K^{m-1} + K^{m-2} + \dots + K^n) |x_1 - x_0| \leq K^n (1 + K + K^2 + \dots) |x_1 - x_0|$$
$$\leq \frac{K^n}{1-K} |x_1 - x_0|$$

Since $[x_n]$ is a Cauchy seq theorem follows

Remark: let $\mathcal{S} = \lim x_n$

let $e_{n+1} = x_{n+1} - \mathcal{S} = g(x_n) - g(\mathcal{S}) = g'(x_n)(x_n - \mathcal{S}) = g'(x_n)e_n$

when $x_n < x_n < \mathcal{S} \quad \sim g'(\mathcal{S})e_n$

$\therefore e_{n+1} \sim g'(\mathcal{S})e_n$ error shows that x_n converges linearly to $\mathcal{S}$

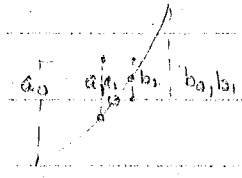
note that $g'(\mathcal{S}) < 1$ by theorem.

a clever pick of g shows $e_{n+1} \sim () e_n^2 \quad h = \frac{1}{f'}$

9 Oct 1972

Review (5 methods)

1. Bisection



$$b_0 - a_0 = M$$

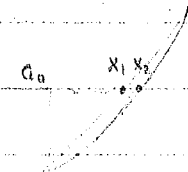
$$|x_{i+1} - x_i| = \frac{M}{2^{i+2}} < \epsilon$$

$$2^{-i} < \epsilon/M$$

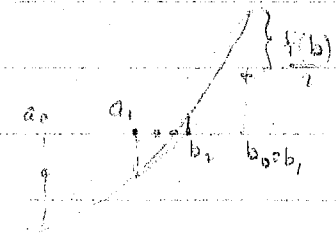
$$-i \log 2 < 2 \log 2 + \log(\epsilon/M)$$

$$i > -2 - \frac{\log(\epsilon/M)}{\log 2}$$

2. Regula Falsi



3. Modified Regula Falsi



4. Secant Method

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

5. Newton

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

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

SUBROUTINE NEWTON(F, FDER, X0, XTOL, FTOL, NTOL, IFLAG)

IFLAG = 0

FdFDER are fn subroutine

DO 20 I = 1, NTOL

FX0 = F(X0)

IF (ABS(FX0), LE. FTOL) RETURN

* IFLAG = 1

DERIV = FDER(X0)

RETURN

IF (ABS(DERIV), LE. FTOL) GO TO #

X0 = X0 - FX0/DERIV

IF (ABS(FX0/DERIV), LE. XTOL) RETURN

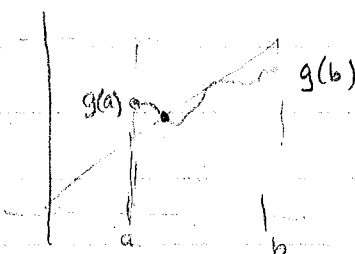
20 CONTINUE

Function Iteration

$G(x) = g$ is needed to be solved $x_{n+1} = g(x_n)$ x_0 given

1) If $x \in [a, b]$ then $g(x) \in [a, b]$

2) a $|g'(x)| \leq K < 1$ $x \in [a, b]$



2b $|g(x_1) - g(x_2)| \leq K |x_1 - x_2|$

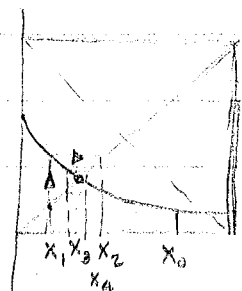
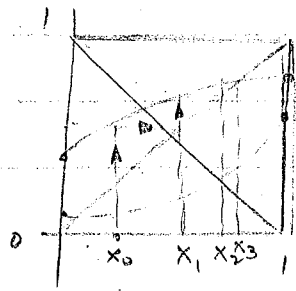
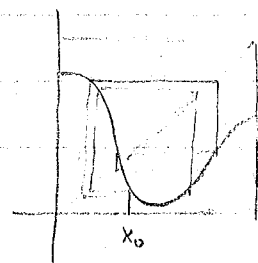
2b $\rightarrow$ 2a because $g(x_1) - g(x_2) = g'(\eta)(x_1 - x_2)$ $x_1 \leq \eta \leq x_2$

by mean value theorem.

$|g(x_1) - g(x_2)| = |g'(\eta)| |x_1 - x_2| \leq K |x_1 - x_2|$

If 1 & 2 hold then $\{x_n\}$ converges for any $x_0 \in [a, b]$ to the unique

$s \in [a, b]$ s.t. $s = g(s)$



Error in functional Iteration

$$e_{n+1} = s - x_{n+1} = g(s) - g(x_n) = g'(\eta_n)(s - x_n) = g'(\eta_n)e_n$$

$\eta_n \in [s, x_n]$ $\therefore$ Since $x_n \rightarrow s$, $\eta_n \rightarrow s$ $e_{n+1} \sim g'(s)e_n$

x_n converges linearly to s whenever $e_{n+1} \sim g'(s) e_n$ holds

Suppose $g'(s) = 0$ then M.V.T.

$$e_{n+1} = \frac{1}{2} g''(\eta_n) e_n^2 \quad \text{where } \eta_n \in [x_n, s]$$

$$e_{n+1} \sim \frac{1}{2} g''(s) e_n^2 \quad \text{quadratic error}$$

Verify that Newton's method is quadratic convergent

$$x = g(x) = x - \frac{f(x)}{f'(x)}$$

$$g'(s) = 1 - \frac{f'(s)}{f'(s)} + \frac{f(s)f''(s)}{f'(s)^2} = 0$$

Secant Method Error

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)} \quad \begin{matrix} f(x_n) - f(x_{n-1}) \\ x_n - x_{n-1} \end{matrix}$$

$$s - x_{n+1} \sim \text{const } e_n e_{n-1} \quad (1+\sqrt{5})/2 \\ = e_{n+1} \sim \text{const } e_n^\alpha \quad 1 < \alpha < 2$$

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

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

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

$$= -e_n e_{n-1} \left[\frac{f(x_n) - f(s)}{x_n - s} - \frac{f(x_{n-1}) - f(s)}{x_{n-1} - s} \right] \\ = \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$$

$$\text{Assume } x_n \rightarrow s \quad = e_n e_{n-1} \frac{\frac{1}{2} f''}{f'} \quad , \quad \text{const} = \frac{\frac{1}{2} f''(s)}{f'(s)}$$

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

$$a_{n+1} = () a_n + () a_{n+1}$$

Isaacs & Keller

Atkinson's Acceleration Method. (Delta²)

$$S - x_{n+1} = g(S) - g(x_n) \quad x_{n+1} = g(x_n) \quad x_1, x_2, \dots, x_i$$

$$g'(\eta_n) [S - x_n]$$

$$S [1 - g'(\eta_n)] = x_{n+1} - g'(\eta_n) x_n$$

$$= (x_{n+1} - x_n) + (x_n [1 - g'(\eta_n)]) \quad \eta_n \in (S, x_n)$$

$$S = x_n + \frac{x_{n+1} - x_n}{1 - g'(\eta_n)}$$

$$g'(\eta_n) \approx \frac{g(x_n) - g(x_{n-1})}{x_n - x_{n-1}} = \frac{x_{n+1} - x_n}{x_n - x_{n-1}} = r_n$$

Define $\hat{x}_n = x_n + \frac{x_{n+1} - x_n}{1 - r_n} = x_{n+1} + \frac{x_{n+1} - x_n}{r_n - 1}$

Theorem: If $x_n \rightarrow S$ then $\hat{x}_n = S + o(S - x_n)$

$$\Delta x_k = x_{k+1} - x_k$$

$$\Delta^2 x_k = \Delta(\Delta x_k) = \Delta(x_{k+1}) - \Delta x_k = x_{k+1} - 2x_{k+1} + x_k$$

$$\hat{x}_n = x_{n+1} - \frac{(\Delta x_n)^2}{\Delta^2 x_{n-1}}$$

Stephenson's Iteration Proc to be programmed

$$g(x) = x$$

Given: x_0 and $g(x)$ $y_0 = x_0$

for $n=0, 1, 2, \dots$ calculate

$$1. \quad S_0 = y_n$$

$$2. \quad S_1 = g(S_0), \quad S_2 = g(S_1)$$

$$3. \quad d = \Delta S_1 = S_2 - S_1, \quad \Delta S_0 = S_1 - S_0$$

$$r = \frac{\Delta S_0}{\Delta S_1} = \frac{S_1 - S_0}{S_2 - S_1}$$

Let $y_{n+1} = S_2 + \frac{d}{r-1}$
 repeat

program Prob. 2.4.3
 $g(x) = \frac{1}{2} + \frac{1}{3} \pi x$

p. 43 2-2.3 part (a) 2-3.1, 5.3
 Study $x = \tan x$; $x^n + \dots + k = 0$ find roots as k varies

2.3-8 $f(x) = e^x - 1 - x - \frac{x^2}{2}$
 $= 1 + \sum \frac{x^n}{n!} - 1 - x - \frac{x^2}{2} = \sum_{n=3}^{\infty} \frac{x^n}{n!}$

$$e^x = 1 + x + \frac{x^2}{2} + \frac{1}{2!} \int_0^x (x-s)^2 e^s ds \quad (uv = uv - \int v du)$$

$$+ \frac{1}{2!} \left[(x-s)^2 e^s \Big|_0^x + 2 \int_0^x (x-s) e^s ds \right]$$

$$+ \frac{1}{2!} \left[-x^2 + 2(x-s)e^s \Big|_0^x + 2 \int_0^x e^s ds \right]$$

$$e^x = 1 + x + \frac{x^2}{2} + \frac{e^s (x)^3}{3!}$$

$$e^x - 1 - x - \frac{x^2}{2} = \frac{e^s (x)^3}{3!} \quad 0 < s < x$$

$\frac{e^s}{3!} \neq 0 \quad \therefore \text{if } f(x) = 0 \quad \underline{x=0}$

23 Oct 1972

1. Polynomial Evaluation
2. Newton's Method for Poly.
3. Fortran of (2)
4. Muller's Method
5. Problem
6. Non-linear systems of Equns (Newton's Method)

Given

$$p(x) = a_0 + a_1x + a_2x^2 + \dots + a_nx^n$$

$$\begin{aligned} p(x) &= a_0 + x(a_1 + a_2x + \dots + a_{n-1}x^{n-2} + a_nx^{n-1}) \\ &= a_0 + x(a_1 + x(a_2 + x(\dots + x(a_{n-1} + x(a_n)) \dots)) \end{aligned}$$

To evaluate $p(x)$ at $x=z$, Set $b_n = a_n$

Do $k = n-1, n-2, \dots, 0$

$$b_k = a_k + z b_{k+1} \quad (*)$$

Then $b_0 = p(z)$

Note that If we define $q(x) = b_1 + b_2x + b_3x^2 + \dots + b_nx^{n-1}$

then $p(x) = q(x)(x-z) + p(z)$ ($p(z) = b_0$)

$q(x)$ is the quotient of p divided by $x-z$ $p(z)$ is remain

Proof:

$$\begin{aligned} a_0 + a_1x + a_2x^2 + \dots + a_nx^n &= (b_1 + b_2x + \dots + b_nx^{n-1})(x-z) + b_0 \\ &= b_0 + b_1x + b_2x^2 + \dots + b_{n-1}x^{n-1} + b_nx^n \\ &\quad - b_1z - b_2zx - b_3zx^2 - \dots - b_nzx^{n-1} \\ &= (b_0 - b_1z) + (b_1 - b_2z)x + \dots + (b_{n-1} - b_nz)x^{n-1} \\ &\quad + b_nx^n \end{aligned}$$

because of $(*)$ Q.E.D.

Note that $p'(z) = q(z)$

because $p'(x) = q(x) + q'(x)(x-z)$; let $x=z$

$$p'(z) = q(z)$$

2. Newton's Method for polynomials

let x_0 and coeff $a_0, a_1, \dots, a_n$ be given

Then: Do for $m=0, 1, 2, \dots$ iteration scheme

set $z = x_m$, $b_n = a_n$, $c_n = a_n$

Do for $k = n-1, n-2, \dots, 1$

$$b_k = a_k + z b_{k+1}$$

$$c_k = b_k + z c_{k+1} \quad \text{defines } c_1 = p'(z)$$

$$\text{set } b_0 = a_0 + z b_1, \quad x_{n+1} = z - b_0/c_1$$

3. DIMENSION $A(N)$

C READ (Array A, M = # of iterations, X_0 initial guess)

$$X = X_0$$

DO 10 I = 1, M

$$B = A(N)$$

$$C = A(N)$$

DO 100 K = 1, N

$$KK = N - K + 1$$

$$B = A(KK) + X * B$$

$$100 \quad C = B + X * C$$

$$B = A(1) + X * B$$

$$\Delta X = B/C$$

IF ($|DELTX| < 10^{-7}$.AND. $|B| < 10^{-7}$) RETURN

$$10 \quad X = X - DELTX$$

Disadvantages of Newton's Method:

1. X_0 must be very good
2. Problems arise for high degree polynomials.
3. Behavior near a double root.

$$p(x) = (x-s)^2 q(x) \quad q(s) \neq 0$$

$$\text{Newton's method } f(x) = 0 \quad x_{n+1} = g(x_n)$$

$$g(x) = x - \frac{f(x)}{f'(x)} \quad e_n = x_n - x_{n-1}$$

Notice $e_{n+1} = g(x_n) - g(x_{n-1}) \sim g'(s)e_n + \frac{1}{2}g''(s)e_n^2$
 for Newton's method.

Note that $g'(x) = 1 - 1 + \frac{f(x)f''(x)}{(f'(x))^2} \Big|_{x=s} = 0$

If $f'(s) = 0$ but $f''(s) \neq 0$ then

Newton's Method converges to s (for sufficient clever x_0)

but $e_{n+1} \sim g'(s)e_n$ this is not zero now.

to see this we must study

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

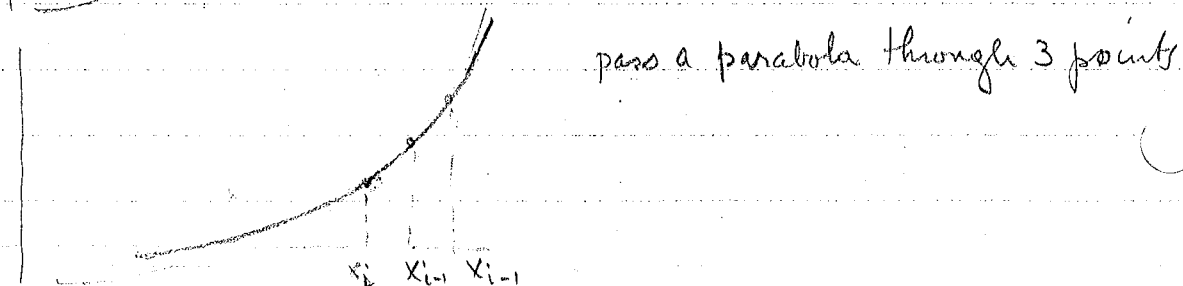
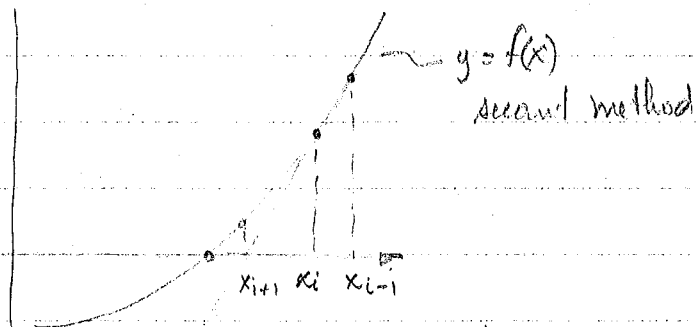
then $e_{n+1} \sim \frac{1}{2}e_n$ when $f'(s) = 0$ & $g(x) = x - f/f'$

then make quadratic convergence by changing g

$$\tilde{g}(x) = x - \frac{2f}{f'}$$

then $e_{n+1} \sim e_n^2$ i.e. $\tilde{g}'(s) = 0$

Muller's Method:



suppose x_i, x_{i-1}, x_{i-2} are known

$$f(x_i), f(x_{i-1}), f(x_{i-2}) = f_i, f_{i-1}, f_{i-2}$$

$$f[x_i, x_{i-1}] = \frac{f_i - f_{i-1}}{x_i - x_{i-1}}$$

$$f[x_i, x_{i-1}, x_{i-2}] = \frac{f[x_i, x_{i-1}] - f[x_{i-1}, x_{i-2}]}{x_i - x_{i-2}}$$

The parabola with eq $y = p(x)$ that passes through (x_i, f_i)
 $(x_{i-1}, f_{i-1}), (x_{i-2}, f_{i-2})$ is given by

$$p(x) = f_i + f[x_i, x_{i-1}](x - x_i) + f[x_i, x_{i-1}, x_{i-2}](x - x_i)(x - x_{i-1})$$

$$p(x_i) = f_i \quad p(x_{i-1}) = f_i - f_i + f_{i-1} = f_{i-1}$$

$$p(x_{i-2}) = f_i + f[x_i, x_{i-1}](x_{i-2} - x_i) + \frac{f[x_i, x_{i-1}] - f[x_{i-1}, x_{i-2}]}{x_i - x_{i-2}} \times [x_{i-2} - x_i][x_{i-2} - x_{i-1}]$$

Write: $p(x) = a_2 x^2 + a_1 x + a_0$

$$a_2 = f[x_i, x_{i-1}, x_{i-2}]$$

$$a_1 = f[x_i, x_{i-1}] - (x_i + x_{i-1}) f[x_i, x_{i-1}, x_{i-2}]$$

$$a_0 = -f[x_i, x_{i-1}]x_i + f[, ,](x_i x_{i-1})$$

Set $x_{i+1} = \text{root of } p(x) = 0$

$$x_{i+1} = \frac{-a_1 \pm \sqrt{a_1^2 - 4a_2 a_0}}{2a_2} = \frac{2a_0}{-a_1 \mp \sqrt{a_1^2 - 2a_2 a_0}}$$

Sign rule: If $a_1 > 0$ take $-$ produces next iterates closer to root
 If $a_1 < 0$ take $+$



Problem:

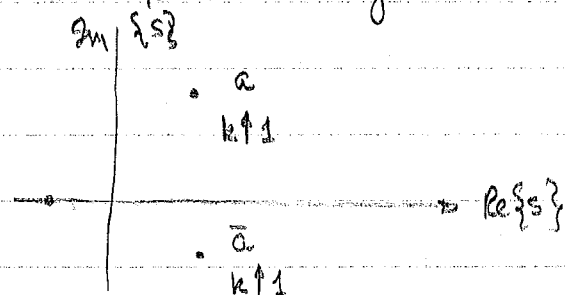
let $0 \leq k \leq 1$ let $g(s) = (s-a)(s-\bar{a})(s-b)$

where $a = 2i+1$ $\bar{a} = 1-2i$ $b = -\frac{1}{2}$

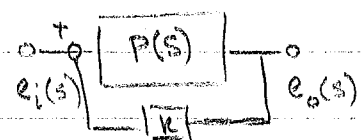
let $G(s, k) = g(s) + k$ find the roots $G(s, k) = 0$

for $0 \leq k \leq 1$ $k = 0, 1, 2, \dots, 1$

what is the smallest k for which $\text{Re}\{s(k)\} \leq 0$?



Use Muller's Method.
& possibly deflation.



$$s = \lambda + i\omega$$

$E_i(t)$ = input signal

$$E_o(t) = \text{output signal} = \int_0^t P(t-\tau) E_i(\tau) d\tau$$

$$e(s) = \int_0^\infty E(t) e^{-st} dt \quad \text{then} \quad e_o(s) = p(s) e_i(s)$$

$$P(s) = \int_0^\infty P(t) e^{-st} dt$$

$$\text{let } p(s) = \frac{1}{(s-a)(s-\bar{a})(s-b)}$$

with feedback $\frac{e_o}{e_i} = \frac{p(s)}{1+kp(s)}$ find roots of $\frac{1}{p(s)} + k$

Algorithm of Muller's Method:

x_i, x_{i-1}, x_{i-2} are given

1. Compute f_i, f_{i-1}, f_{i-2}

2. Compute $h_i = x_i - x_{i-1}$ and $\lambda_i = \frac{h_i}{x_{i-1} - x_{i-2}}$

$$x_{i+1} = x_i + h_i \lambda_{i+1}$$

$$= x_i + \frac{(x_i - x_{i-1})(x_{i+1} - x_i)}{x_i - x_{i-1}} = x_{i+1}$$

$$3. \quad g_i = (1 + 2\lambda_i)(f_i - f_{i-1}) - \lambda_i^2(f_{i-1} - f_{i-2})$$

$$4. \quad \lambda_{i+1} = \text{search} \quad -(\quad) \quad \gamma$$

Choose $\pm$ so that λ_{i+1} is min

$$5. \quad x_{i+1} = x_i + h_i \lambda_{i+1}, \quad h_{i+1} = h_i \lambda_{i+1}, \quad f_{i+1} = f(x_{i+1})$$

6. Increase i by 1 and go to 3 or return

System of Nonlin Eq

$$f(x, y), g(x, y) = 0 \quad \text{find roots}$$

$$\text{write } F(x, y) = x$$

$$G(x, y) = y$$

$$\text{write } \underline{F}(\underline{x}) = \underline{x}$$

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

$$\underline{f}'(\underline{x}) = \begin{pmatrix} f_x & f_y \\ g_x & g_y \end{pmatrix}$$

$$\underline{x}_{n+1} = \underline{F}(\underline{x}_n)$$

$$\underline{F}(\underline{x}) = \underline{x} - [\underline{f}'(\underline{x})]^{-1} \underline{f}(\underline{x})$$

10-30-72

1. Newton's Method for nonlinear systems.

2. Review of Linear Algebra.

3. Num. Soln for linear systems

(a) Back Substitution

(b) Gaussian Elimination

$$x_{i+1} = x_i + \lambda_i h_i$$

$$h_i = x_i - x_{i-1}$$

problem: $\frac{1}{2} \pm 2i$ instead of $1 \pm 2i$

$$\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \quad \underline{F}(\underline{x}) = \begin{pmatrix} F_1(x_1, x_2, \dots, x_n) \\ F_2(x_1, \dots, x_n) \\ \vdots \\ F_n(x_1, \dots, x_n) \end{pmatrix}$$

Problem find solns of $\underline{f}(\underline{x}) = 0$

$$\begin{aligned} f_1(x_1, \dots, x_n) &= 0 \\ f_2(x_1, \dots, x_n) &= 0 \\ &\vdots \\ f_n(x_1, \dots, x_n) &= 0 \end{aligned}$$

Theorem: let $G \subset \mathbb{R}^N$ (closed subset)

& suppose $F: G \rightarrow G$.

$$\left. \begin{aligned} & \sum_{n=0}^{\infty} F^n(x) \text{ for any } \underline{x}_1, \underline{x}_2 \in G \end{aligned} \right\}$$

$$\|F(\underline{x}_1) - F(\underline{x}_2)\| \leq K \|\underline{x}_1 - \underline{x}_2\|, \quad K < 1$$

then for any $\underline{x}_0$ in G the sequence of vectors $\underline{x}_n$ where $\underline{x}_{n+1} = F(\underline{x}_n)$ converges to a unique vector $\underline{\xi}$ i.e. $\|\underline{x}_n - \underline{\xi}\| \rightarrow 0$ as $n \rightarrow \infty$

& $\underline{\xi} = F(\underline{\xi})$ prove:

def: norm $\|\cdot\|$ if $\underline{x} \in \mathbb{R}^N$ $\|\underline{x}\|_{\infty} = \max_{1 \leq i \leq n} |x_i|$

Error analysis of fixed point iteration

$$\begin{aligned} \underline{e}^n &\equiv \underline{x}^n - \underline{x}^{n-1} & \underline{e}^{n+1} &\equiv \underline{x}^{n+1} - \underline{x}^n = F(\underline{x}^n) - F(\underline{x}^{n-1}) = \\ & & &= F'(\eta^n) [\underline{x}^n - \underline{x}^{n-1}] \end{aligned}$$

$$F'(\underline{x}) = \left[\frac{\partial F_i(\underline{x})}{\partial x_j} \right]$$

$$e_i^{n+1} = \sum_{j=1}^n \frac{\partial F_i}{\partial x_j}(\eta^n) e_j^n$$

η^n is on the line connecting $\underline{x}^n$ & $\underline{x}^{n-1}$

$\therefore \underline{e}^{n+1} \approx F'(\underline{\xi}) \underline{e}^n$ $F'(\underline{\xi})$ has "norm" less than 1

error is first order unless $F'(\underline{\xi}) = 0$

to have $\underline{F}(\underline{x}) = 0$ choose $\underline{x} = \underline{F}(\underline{x}) = \underline{x} - [\underline{f}'(\underline{x})]^{-1} \underline{f}(\underline{x})$

Recall $\underline{f}(\underline{x}) = 0$ when $\underline{x} = \underline{x}$

Apply this to two eqns. $\left. \begin{array}{l} f(x,y)=0 \\ g(x,y)=0 \end{array} \right\} \underline{f}(\underline{x})=0$
 $\left. \begin{array}{l} F(x,y)=x \\ G(x,y)=y \end{array} \right\} \underline{F}(\underline{x})=\underline{x}$

$$\underline{f}'(\underline{x}) = \begin{bmatrix} f_x & f_y \\ g_x & g_y \end{bmatrix} \quad [\underline{f}'(\underline{x})]^{-1} = \frac{1}{f_x g_y - f_y g_x} \begin{bmatrix} g_y & -f_y \\ -g_x & f_x \end{bmatrix}$$

$$\begin{pmatrix} F(x,y) \\ G(x,y) \end{pmatrix} = \begin{pmatrix} x \\ y \end{pmatrix} - \frac{1}{f_x g_y - f_y g_x} \begin{bmatrix} g_y & -f_y \\ -g_x & f_x \end{bmatrix} \begin{bmatrix} f \\ g \end{bmatrix}$$

Newton's Method (x_0, y_0) $x_{n+1} = F(x_n, y_n)$
 $y_{n+1} = G(x_n, y_n)$

Example $x^2 + xy^3 - 9 = 0$
 $3x^2y - y^3 - 4 = 0$

Derivative Matrix:

$$A = \begin{bmatrix} 2x + y^3 & 3xy^2 \\ 6xy & 3x^2 - 3y^2 \end{bmatrix} \quad \det A = [3(x^2 - y^2)(2x + y^3) - 18x^2y^3] = J(x,y)$$

$$F(x,y) = x - \frac{(x^2 + xy^3 - 9)(3x^2 - 3y^2) - (3x^2y - y^3 - 4)(3xy^2)}{J}$$

$$G(x,y) = y - \frac{\dots}{J}$$

Error analysis for N.M. in $\mathbb{R}^N$ is identical to $\mathbb{R}^1$. & same general remarks hold.

Review of L.A.

$\underline{x}, \underline{y}, \dots, \underline{a}, \underline{b}$ vectors

$$A\underline{x} = \underline{b} \quad \sum_{j=1}^n a_{ij} x_j = b_i$$

$A, B, \dots$ matrices

$$A = (a_{ij}) \quad \underline{x} = (x_i)$$

1. Vectors form a vector space, i.e. $\mathbb{R}^N$ is a vector space.

$$(i) \quad \underline{x} + \underline{y} \in \mathbb{R}^N$$

$$\underline{x} + \underline{y} = \underline{y} + \underline{x}$$

$$(ii) \quad \alpha \underline{x} \in \mathbb{R}^N$$

$$0 \text{ is the additive identity } \underline{0} + \underline{x} = \underline{x}$$

2. $n \times n$ matrices form a vector space

$$(i) \quad A + B \in V$$

$$(ii) \quad \alpha A \in V$$

$$0 \text{ matrix is } \underline{0}, \quad A + \underline{0} = A$$

3. Matrices have a mult. rule where

$$(AB)_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

the mult. satisfies $A(B+C) = AB + AC$

$$A(BC) = (AB)C$$

non-commutative. $AB \neq BA$ is general $I = \delta_{ij} \begin{pmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{pmatrix}$

$$BI = B \quad IA = A$$

Linear Dependence

A set of vectors $\underline{x}_1, \underline{x}_2, \dots, \underline{x}_k$ are linear indep. iff

$$\text{for } \alpha_1 \underline{x}_1 + \alpha_2 \underline{x}_2 + \dots + \alpha_n \underline{x}_n = \underline{0} \quad ; \quad \alpha_1 = \alpha_2 = \alpha_3 = \dots = \alpha_n = 0$$

Let A be $n \times n$ matrix & let $\underline{a}_1, \underline{a}_2, \dots, \underline{a}_n$ be its column vectors

Def: The largest number of L.I. $\underline{a}$'s is called the rank of A .

Thm: If $\text{rank } A = r$ then there are exactly r l.i.

row vectors of A . Hence $\text{rank}(A) = \text{column rank}(A) = \text{row rank}(A)$

Let $A = (a_{ij})$. Suppose, without loss of generality that the first r columns are L.I. then if $\sum_{i=1}^n c_i a_{ij} = 0$ then $c_1 = c_2 = \dots = c_r = 0$.

$j = 1, 2, \dots, n$

Let d_j be some constants & add $\sum_{j=1}^n \sum_{i=1}^n d_j c_i a_{ij} = 0$

$\sum_{i=1}^n c_i \left(\sum_{j=1}^n d_j a_{ij} \right) = 0$; To deduce that if $\sum d_j a_{ij} = 0$ then it cannot be that $d_1 = d_2 = \dots = d_n = 0$ if it were zero.

Problem: prove the above.

We wish to solve the linear system $AX = b$

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$\vdots$$

$$a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m$$

Given A & b does $\exists$ a soln?

① Alternative theorem: If $AX = 0 \iff X = 0$ then $AX = b$ has a unique soln. for every b & $\text{rank}(A) = n$. column vectors of A are L.I.

② Otherwise $\exists$ vectors $x^{(1)}, x^{(2)}, x^{(3)}, \dots, x^{(p)}$ s.t. if $AX = 0$ then $X = \sum_{j=1}^p \alpha_j x^{(j)}$ $n \geq p \geq 1$

In this case $AX = b$ is not ^{uniquely} solvable for every b .

$AX = b$ is solvable if $\text{rank}(A) = \text{rank}(A, b) < n$

where A, b is the set of column vectors of A & the column vector b .

Numerical Treatment

1) Back Substitution Algorithm

If A is upper triangular 

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$+ a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$$\dots$$

$$a_{n-1,n-1}x_{n-1} + a_{n-1,n}x_n = b_{n-1}$$

$$a_{nn}x_n = b_n$$

Assume $a_{kk} \neq 0$

$$x_n = \frac{b_n}{a_{nn}}, \quad x_{n-1} = \frac{b_{n-1} - a_{n-1,n}x_n}{a_{n-1,n-1}}, \dots$$

For $k = n, \dots, 1$ 

$$x_k = \frac{b_k - \sum_{j=k+1}^n a_{kj}x_j}{a_{kk}}$$

where $\sum_{j=n+1}^n$ is interpreted as zero.

2) Gaussian Elimination

a) Assume $a_{11} \neq 0$

mult $2, 3, \dots, n$ by $\frac{a_{11}}{a_{j1}} \quad j=2, 3, \dots, n$

$\pm$ subtract from 1st. eq

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{22}^{(1)}x_2 + \dots + a_{2n}^{(1)}x_n = b_2^{(1)}$$

$$a_{32}^{(1)}x_2 + \dots + a_{3n}^{(1)}x_n = b_3^{(1)}$$

$$\vdots$$

$$a_{n2}^{(1)}x_2 + \dots + a_{nn}^{(1)}x_n = b_n^{(1)}$$

$$a_{2j}^{(1)} = a_{1j} - \frac{a_{11}}{a_{21}} a_{2j}$$

$$b_2^{(1)} = b_1 - \frac{a_{11}}{a_{21}} b_2$$

for $i=2, 3, 4, \dots$

$$a_{ij}^{(1)} = a_{ij} - \frac{a_{i1}}{a_{11}} a_{1j}$$

$$b_i^{(1)} = b_i - \frac{a_{i1}}{a_{11}} b_1$$

Assume $a_{11}^{(1)} \neq 0$

Next G.E. with partial pivoting & retaining of multipliers

11/6/72

1. Row & Column Rank
2. Gauss Elimination
3. Pivoting
4. Form of Gauss
5. Triangular Factorization Substitution Algorithm
6. Form for Subst. Algo
7. Norms for vectors & matrices

If $A = (a_{ij})$ $a_1, a_2, \dots, a_n$ are its columns

then c-rank $(A) = \text{largest \# of L.I. c-vectors}$

if A has c-rank $(A) = n$ then row rank $(A) = n$. also.

Proof: Hypothesis If $\sum_{j=1}^n a_{ij} x_j = 0 \quad i=1, 2, \dots, n$ then $x_1 = x_2 = \dots = x_n = 0$

To show that row rank $= n$ i.e. If $\sum_{i=1}^n a_{ij} y_i = 0^{\otimes} \quad y_1 = y_2 = \dots = y_n = 0$

Suppose $\otimes$ holds but some $y_k \neq 0$ then $\sum_{i=1}^n y_i \left(\sum_{j=1}^n a_{ij} x_j \right) = 0$

$\Rightarrow a_{kj} = 0 \quad j=1, \dots, n \Rightarrow$ the k^{th} entry of the column vectors $a_1, a_2, \dots, a_n$ is zero $\Rightarrow$ the vectors are L. Dependent.

Gauss Elimination $AX=b$

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 3 & -1 & 1 \\ 4 & 0 & 2 \end{pmatrix} \quad b = \begin{pmatrix} 2 \\ 3 \\ 4 \end{pmatrix}$$

No pivoting

$$W = \left(\begin{array}{ccc|c} 1 & 2 & 3 & 2 \\ 3 & -1 & 1 & 3 \\ 4 & 0 & 2 & 4 \end{array} \right) \rightarrow \left(\begin{array}{ccc|c} 1 & 2 & 3 & 2 \\ 0 & -7 & -8 & -3 \\ 0 & -8 & -10 & -4 \end{array} \right) \rightarrow \left(\begin{array}{ccc|c} 1 & 2 & 3 & 2 \\ 0 & -7 & -8 & -3 \\ 0 & 8/7 & -6/7 & -4/7 \end{array} \right)$$

$$AX=b \Leftrightarrow UX=\tilde{b}$$

use back Subst. now to solve triang matrix

Pivoting uses the idea of non division by a small no.

Let $A = (a_{ij})$

$d_i = \max_{1 \leq j \leq n} |a_{ij}|$ to eliminate x_i choose that row p_i for pivot for

$$\text{which } \frac{|a_{p_i,1}|}{d_{p_i}} \geq \frac{|a_{i1}|}{d_i} \quad i = p_2, \dots, p_n$$

compute $\frac{|a_{i1}|}{d_i}$ find row $\Rightarrow \max_{1 \leq i \leq n} \frac{|a_{i1}|}{d_i}$ & interchange rows

do this for all matrices.

Outline of Algorithm (without maximal pivoting)

1. Initialize the pivot vector $p = (1, 2, \dots, n)$

and $W = A$ initially

2. DO $k=1, \dots, n-1$

search for pivot p_j $j \geq k \Rightarrow W_{p_j, k} \neq 0$

Exchange contents of row p_k & row p_j

DO $I = k+1, \dots, n$ $m = \frac{W_{p_i, k}}{W_{p_k, k}}, W_{p_i, k} = \frac{W_{p_i, k}}{W_{p_k, k}}$

DO $J = k+1, \dots, n$

$$W_{p_i, j} = W_{p_i, j} - m W_{p_k, j}$$

C Gauss Elimination N-dim

DIMENSION A(N,N), W(N,N), IPVOT(N), D(N)

IFLAG=1

C INITIALIZE IPVOT, D

DO 10 I=1,N

IPVOT(I)=I

ROWMAX=0.

DO 9 J=1,N

IF (ABS(W(I,J)), LE. ROWMAX) GO TO 9

ROWMAX = ABS(W(I,J))

9 CONTINUE

IF (ROWMAX, LE. 1.0E-07) GO TO 999

10 D(I)=ROWMAX.

C BEGIN ELIMINATION

NM1 = N-1

DO 20 K=1, NM1

J=K

KP1 = K+1

IP = IPVOT(K)

COLMAX = ABS(W(IP,K)) / D(IP)

DO 11 I=KP1, N

IP = IPVOT(I)

AW = ABS(W(IP,K)) / D(IP)

IF (AW, LE. COLMAX) GO TO 11

COLMAX = AW

J=I

11 CONTINUE

C Exchange of indices

IPK = IPVOT(J)

IPVOT(J) = IPVOT(K)

IPVOT(K) = IPK

C LAST 2 STEPS IN ELIM.

DO 20 I=KP1, N

IP = IPVOT(I)

W(IP,K) = W(IP,K) / W(IPK,K)

RATIO = -W(IP,K)

DO 20 J=KP1, N

W(IP,J) = W(IP,J) + RATIO * W(IPK,J)

20 CONTINUE

RETURN

999 IFLAG=2

RETURN

Problem on Linear Equations

Let

$$H_n = \begin{bmatrix} 1 & 1/2 & 1/3 & 1/4 & \dots & 1/n \\ 1/2 & & & & & \\ 1/3 & & & & & \\ \vdots & & & & & \\ 1/n & 1/(n+1) & & & & 1/(n-1) \end{bmatrix}$$

For $n=5$ DO Gaussian Elimination to triangular form.

Find inverse of matrix.

Solve also by hand

5) Forward & Backward Substitution

Backward Subst. clear i.e. how to solve when system is in triangular form.

$$Ax = b$$

$$W = [A, b] \quad \text{if no pivoting}$$

$$\begin{pmatrix} U \\ M \end{pmatrix} \begin{pmatrix} x \end{pmatrix} = \begin{pmatrix} \tilde{b} \end{pmatrix}$$

$$\begin{pmatrix} 2 & 3 & 1 \\ 3 & 1 & 0 \\ -1 & -2 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \\ 5 \end{pmatrix} \quad W = \left(\begin{array}{ccc|ccc} 2 & 3 & 1 & 3 & & \\ 3 & 1 & 0 & 2 & & \\ -1 & -2 & 0 & 5 & & \end{array} \right) \quad b^{(0)} = b$$

$$\begin{pmatrix} 2 & 3 & 1 & 3 \\ 3/2 & x & x & 2 - 9/2 \\ -1/2 & -1/2 & x & 5 + 1/3 \end{pmatrix} \quad \begin{aligned} b_1^{(1)} &= b_1^{(0)} \\ b_2^{(1)} &= b_2^{(0)} - m_{21} b_1^{(0)} \\ b_3^{(1)} &= b_3^{(0)} - m_{31} b_1^{(0)} \end{aligned}$$

$$\begin{pmatrix} \end{pmatrix} \quad \begin{aligned} b_1^{(2)} &= b_1^{(1)} \\ b_2^{(2)} &= b_2^{(1)} \\ b_3^{(2)} &= b_3^{(1)} - m_{32} b_2^{(1)} \end{aligned}$$

$$b_1^{(2)} = b_1^{(0)}$$

$$b_2^{(2)} = b_2^{(0)} - m_{21} b_1^{(0)} = b_2^{(1)}$$

$$b_3^{(2)} = b_3^{(0)} - m_{31} b_1^{(0)} - m_{32} b_2^{(1)}$$

Given

$$\begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}$$

$Ax = b$ the $\tilde{b}_k$ in $Ux = \tilde{b}$

$$\tilde{b}_k = b_k - \sum_{j=1}^{k-1} m_{kj} \tilde{b}_j \quad k=1, 2, \dots, n$$

What if we had pivoting? The IPivot vector $\begin{pmatrix} p_1 \\ p_2 \end{pmatrix}$ is stored. Recall

$$m_{kj} = W_{kj} \text{ in routine } k > j. \text{ In general then } \tilde{b}_k = b_{p_k} - \sum_{j=1}^{k-1} W_{p_k j} \tilde{b}_j$$

Fortran program Assume for forward & backward subst:

$$\text{Backward Subst. } X_k = \tilde{b}_n - \sum_{j=k+1}^n W_{p_k j} X_j \quad k=n, n-1, \dots, 1$$

DIM N(N,N) B(N) X(N), IPivot(N), BTL(N) N ≥ 2

IP = IPivot(1) SUM = 0.

BTL(1) = B(IP) DO 14 J = KKPI, N

DO 11 K = 2, N 14 SUM = SUM + W(IP, J) * X(J)

KMI = K-1 13. X(KK) = (BTL(KK) - SUM) / W(IP, K)

SUM = 0.0 RETURN

IP = IPivot(K)

DO 12 J = 1, KMI

12 SUM = SUM + W(IP, J) * BTL(J)

11 BTL(K) = B(IP) - SUM

X(N) = BTL(N) / W(IPivot(N), N)

DO 13 K = 2, N

KK = N+1-K

KKPI = KK+1

IP = IPivot(KK)

KK = N+1, 1

KKPI = N, 2

Let V be a vector space. denote vector by x, y

$\|x\|$ is a positive number associated with each x called norm of x
with the following $\|x\| \geq 0$ $\|x\| = 0 \iff x = 0$

$$\|\alpha x\| = |\alpha| \|x\|$$

$$\|x + y\| \leq \|x\| + \|y\|$$

Example

$$\|x\|_{\infty} = \max_{1 \leq i \leq n} |x_i|$$

$$\|x\|_1 = \sum_{i=1}^n |x_i|$$

$$\|x\|_2 = \left(\sum_{i=1}^n x_i^2 \right)^{1/2}$$

$$\|x\|_p = \left(\sum_{i=1}^n |x_i|^p \right)^{1/p}$$

$$\|x\|_{p,w} = \left(\sum_{i=1}^n |x_i|^p w_i \right)^{1/p} \quad w_i \geq 0$$

Check these are norms

11-13-72

1. Tridiagonal factoring
2. Error - Residual Norms
3. Condition Number
4. Iterative improvement
5. Iterative Methods.

Recall that Gauss Elimination transforms a given nonsingular matrix A into a matrix U and a pivot vector p .

Let $U = (u_{ij})$ be the matrix $u_{ij} = \begin{cases} w_{p_i, j} & i \leq j \\ 0 & i > j \end{cases}$

U is upper triangular & the system

$$Ax = b \iff UX = \tilde{b} \quad \text{where}$$

$$\tilde{b}_k = b_{p_k} - \sum_{j=1}^{k-1} w_{p_k j} \tilde{b}_j \quad || \quad b_{p_k} = \tilde{b}_k + \sum_{j=1}^{k-1} w_{p_k j} \tilde{b}_j$$

Let L be the matrix defined as follows

$$l_{ij} = \begin{cases} w_{p_i j} & i > j \\ 1 & i = j \\ 0 & i < j \end{cases}$$

Let $\tilde{b} = (b_{p_j})$ permuted b -vector

then $\tilde{b} = L \tilde{b}$

Now from $UX = \tilde{b} \quad LUx = L\tilde{b} = \tilde{b} = P^{-1}b$

where P is a permutation matrix.

$$P \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix} (i) = \begin{pmatrix} 0 \\ \vdots \\ 1 \\ 0 \end{pmatrix} (p_j) \quad \text{ie } P \text{ permutes the rows of } I$$

according to the pivot strategy p .

Note that if P is any perm matrix then $P^2 = I$ i.e. $P = P^{-1}$

From $LUx = P^{-1}b \Rightarrow PLUx = b \Rightarrow PLU = A$

since it is true for $\forall x$.

based on P lower triang upper triang

2) Problem $AX=b$ By G.E. $\hat{x}$ is obtained

Error $x - \hat{x} = e$

Residual $r = Ae = A(x - \hat{x}) = b - A\hat{x}$

Since r is computable how can we use it to estimate e .

if r is small e may be tremendously large

Vector Matrix Norms $x, y, \dots$ will be vectors. By $\|x\|$ we

define a non negative, for each x .

1. $\|x\| \geq 0$; $\|x\| = 0$ iff $x = 0$

2. $\|ax\| = |a| \|x\|$

3. $\|x+y\| \leq \|x\| + \|y\|$

Examples of norms.

$$\|x\|_{\infty} = \max_{1 \leq j \leq n} |x_j|$$

$$\|x\|_{p,w} = \left(\sum_{j=1}^n |x_j|^p w_j \right)^{1/p} \quad w_j \geq 0$$

$$\|x\|_1 = \sum_{j=1}^n |x_j|$$

$$\|x\|_2 = \left(\sum_{j=1}^n x_j^2 \right)^{1/2}$$

Homework: $\lim_{p \rightarrow \infty} \|x\|_p = \|x\|_{\infty}$

Verify 3) $\|x+y\|_{\infty} = \max_{1 \leq j \leq n} |x_j + y_j| \leq \max_{1 \leq j \leq n} \{ |x_j| + |y_j| \} \leq \|x\|_{\infty} + \|y\|_{\infty}$

$$\|x+y\|_1 = \sum_{j=1}^n |x_j + y_j| \leq \sum_{j=1}^n (|x_j| + |y_j|) = \|x\|_1 + \|y\|_1$$

$$\|x+y\|_2 \quad \text{note that if } x \cdot y = \sum_{i=1}^n x_i y_i = x^T y = (x_1, \dots, x_n) \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$\|x\|_2 = \sqrt{x \cdot x} \quad \text{Now consider } \|x+y\|_2^2 = (x+y) \cdot (x+y) \\ = x \cdot x + 2x \cdot y + y \cdot y$$

$$\frac{x \cdot y}{\|x\|_2 \|y\|_2} = \cos(x, y) \leq 1 \quad \therefore \|x \cdot y\| \leq \|x\|_2 \|y\|_2 \quad \text{Schwarz's inequality}$$

$$\|x+y\|_2^2 \leq \|x\|_2^2 + 2\|x\|_2 \|y\|_2 + \|y\|_2^2 \leq (\|x\|_2 + \|y\|_2)^2 \quad \text{QED}$$

Natural (induced) matrix norms: Given a vector norm and a matrix A then

$$\|A\| = \max_{x \neq 0} \frac{\|Ax\|}{\|x\|} = \max_{\|x\|=1} \|Ax\|$$

$\| \cdot \|_\infty$ induces the max norm $\|A\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$

$$\|A\|_1 = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}|$$

$$\|A\|_2 = \sqrt{\rho(A^T A)}$$

what properties does $\|A\|$ have?

1) $\|A\| \geq 0$; $\|A\| = 0 \iff A = 0$

2) $\|aA\| = |a| \|A\|$

3) $\|A+B\| \leq \|A\| + \|B\|$

4) $\|AB\| \leq \|A\| \|B\|$

4) from definition $\|Ax\| \leq \|A\| \|x\|$

$$\|ABx\| \leq \|A\| \|Bx\| \leq \|A\| \|B\| \|x\|$$

$$\max_{x \neq 0} \frac{\|ABx\|}{\|x\|} \leq \|A\| \|B\| = \|AB\|$$

To show $\|A\|_\infty = \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}|$ one must show that $\|Ax\|_\infty \leq \|A\|_\infty \|x\|_\infty$

for every x and that $\exists$ some x for which $=$ holds.

$$\|Ax\|_\infty = \max_{1 \leq i \leq m} \left| \sum_{j=1}^n a_{ij} x_j \right| \leq \max_{1 \leq i \leq m} \sum_{j=1}^n |a_{ij}| |x_j|$$

$$\leq \|x\|_\infty \|A\|_\infty$$

Let i_0 be the index for which $\sum_{j=1}^n |a_{i_0 j}|$ is max.

$$\text{let } x_j = \begin{cases} 1 & a_{i_0 j} \geq 0 \\ -1 & a_{i_0 j} < 0 \end{cases}$$

Then $\|Ax\|_\infty = \max_{1 \leq i \leq m} \left| \sum_{j=1}^n a_{ij} x_j \right| \geq \left| \sum_{j=1}^n a_{i_0 j} x_j \right|$

$$\text{but we have chosen } x_j \therefore \sum_{j=1}^n |a_{i_0 j}| = \|A\|_\infty = \|A\|_\infty \|x\|_\infty$$

$\therefore \|Ax\|_\infty = \|A\|_\infty \|x\|_\infty$ for this vector.

check: $\|A\|_1 = \text{column sum}$.

4) Iterative Improvement.

$AX=b$, $\hat{x}^{(1)}$ computed solution (G.E.)

Compute $r^{(1)} = Ae = b - A\hat{x}^{(1)}$

Then solve $Ae = r^{(1)}$, call $\hat{e}^{(1)}$ the computed solution

Construct $\hat{x}^{(2)} = \hat{x}^{(1)} + \hat{e}^{(1)}$

continue. This will usually provide improvement.

3) Back to Condition number

$$\|Ax\| \leq \|A\| \|x\|$$

$$\frac{\|x\|}{\|A^{-1}\|} \leq \|Ax\|$$

$$\|x\| = \|(A^{-1}Ax)| \leq \|A^{-1}\| \|Ax\|$$

$$\therefore \frac{1}{\|A^{-1}\|} \leq \frac{\|Ax\|}{\|x\|} \leq \|A\|$$

To estimate bounds for $\frac{\|e\|}{\|x\|} = \frac{\|x - \hat{x}\|}{\|x\|}$

$$\frac{\|r\|}{\|A\|} \leq \|A^{-1}r\| \leq \|A^{-1}\| \|r\|$$

$$x = A^{-1}b \quad \frac{\|b\|}{\|A\|} \leq \|x\| \leq \|A^{-1}\| \|b\|$$

$$\frac{1}{\|b\| \|A^{-1}\| \|A\|} \leq \frac{\|e\|}{\|x\|} \leq \frac{\|A^{-1}\| \|r\|}{\|b\|} \cdot \|A\|$$

Let $\text{cond}(A) = \|A\| \|A^{-1}\| \geq 1$

$$\frac{1}{\text{cond}(A) \|b\|} \leq \frac{\|e\|}{\|x\|} \leq \text{cond}(A) \left(\frac{\|r\|}{\|b\|} \right) \text{ relative residual}$$

$$\frac{\|e\|}{\|x\|} \leq \text{cond}(A) \cdot m \cdot 2^{-t} \quad t = \text{digit binary} \quad \text{G.E. only.}$$

Clearly $\|A\|$ is computable; $\|A^{-1}\|$ ^{is not}, let B be any singular matrix & let x be s.t. $Bx=0$

$$\|A-B\| \|x\| \geq \|(A-B)x\| = \|Ax\| \geq \frac{\|x\|}{\|A^{-1}\|}$$

$$\therefore \frac{1}{\|A^{-1}\|} \leq \|A-B\| \quad \text{find a } B \text{ s.t. } \|A-B\| \text{ is very small}$$

5) Iterative Methods

$$Ax=b \Rightarrow (b-Ax)=0 \quad \text{let } C \text{ be nonsingular i.e. } \det C \neq 0$$

$$\Rightarrow C^{-1}(b-Ax)=0$$

$$x + C^{-1}(b-Ax) = x$$

$$(I - C^{-1}A)x + C^{-1}b = x \Rightarrow x = Bx + \bar{c}$$

Fixed pt. scheme where $B = I - C^{-1}A$ & $\bar{c} = C^{-1}b$.

Iterative Scheme

$$x^{(m+1)} = Bx^{(m)} + \bar{c}, \quad x^{(0)} = 0$$

Q: does $x^{(m)}$ converge? how to choose C to do the job "best".

A: $x^{(m)}$ converges $\{ \|x^{(m)} - x\| \xrightarrow{m \rightarrow \infty} 0 \}$ iff $\|B\| < 1$

Find C s.t. C^{-1} exists $\|I - C^{-1}A\| < 1$ C^{-1} must be easily computed

Decompose A into $\hat{L} + \hat{D} + \hat{U}$ lower, diag, upper

Jacoby let $C = \hat{D}$ (assume $a_{ii} \neq 0$)

$$\therefore x_i^{(m+1)} = \sum_{j \neq i} -\frac{a_{ij}}{a_{ii}} x_j^{(m)} + \frac{b_i}{a_{ii}}$$

Jacoby converges when $|a_{ii}| > \sum_{j \neq i} |a_{ij}|$ diagonal dominance

Proof: $\|B\|_\infty = \max_{1 \leq i \leq n} \sum_{j \neq i} \left| \frac{a_{ij}}{a_{ii}} \right| < 1$

2) Gauss-Seidel Choose $\hat{L} + \hat{D}$
 $\|I - (\hat{L} + \hat{D})^{-1}A\| \leq \|I - \hat{D}^{-1}A\|$ not true in general

$$x^{m+1} = x^m + (\hat{L} + \hat{D})^{-1} (b - Ax^m)$$

$$(\hat{L} + \hat{D}) x^{m+1} = (\hat{L} + \hat{D}) x^m + b - Ax^m$$

$$\hat{D} x^{m+1} = -\hat{L} x^{m+1} - \hat{U} x^m + b$$

G.S. $x^{m+1} = -\hat{D}^{-1} \{-\hat{L} x^{m+1} - \hat{U} x^m + b\}$

$$i=1, \dots, n \quad x_i^{(m+1)} = -\frac{1}{a_{ii}} \left[\sum_{j=1}^{i-1} a_{ij} x_j^{m+1} + \sum_{j=i+1}^n a_{ij} x_j^m - b_i \right]$$

Let $A = A^T$ then A has real eigenvalues & n linearly indep eigenvectors. $\Rightarrow$ the k^{th} eigenvector $\cdot j^{\text{th}}$ eigenvector = δ_{kj}

Nov. 20, 1972

1. Linear Algebra (Review)
2. Matrix Norms
3. Convergence of It Schemes (Review)
4. Power method for λ -vector.

$$H_n = \begin{pmatrix} 1 & 1/2 & 1/3 & 1/4 & 1/5 \\ 1/2 & & & & \\ 1/3 & & & & \\ 1/4 & & & & \\ 1/5 & & & & 1/9 \end{pmatrix} \quad H_n^{-1}$$

$$\text{cond}(A) = \|A\|_\infty \|A^{-1}\|_\infty = \|A\|_1 \|A^{-1}\|_1 = 5 \times 10^5$$

$$\det A = \sum_{(\sigma)} \text{sign}(\sigma) a_{1, i_1} a_{2, i_2} \dots a_{n, i_n}$$

$$\sigma = \begin{pmatrix} 1 & 2 & \dots & n \\ i_1 & i_2 & \dots & i_n \end{pmatrix} \quad n! \text{ terms in sum}$$

$$\text{sign}(\sigma) = \begin{cases} +1 & \sigma \text{ even} \\ -1 & \sigma \text{ odd} \end{cases}$$

Theorem: $\det(AB) = \det(A) \det(B)$

Theorem: A is invertible iff $\det(A) \neq 0$

Proof:

$$A = PLU$$

P is a permutation matrix

$$L = \begin{pmatrix} 1 & & 0 \\ x & 1 & \\ & & \ddots \end{pmatrix} \quad U = \begin{pmatrix} u_{11} & & \\ & \ddots & \\ 0 & & u_{nn} \end{pmatrix}$$

$$\det A = \det P \det L \det U$$

$$= \text{sign}(P) \cdot 1 \cdot u_{11} u_{22} \dots u_{nn}$$

if $\exists a \text{ } u_{ii} = 0$ then no back sub is possible $\therefore A^{-1}$ is not invertible

Cramer's Rule: $Ax = b$

$$x_j = \frac{\det A^j}{\det A} \quad \text{where } A^j \text{ has } j^{\text{th}} \text{ replaced by } b$$

Eigenvalues λ of a matrix is a no. real or complex $\Rightarrow$

$$\det(A - \lambda I) = 0$$

$$\therefore \exists x \neq 0, \Rightarrow (A - \lambda I)x = 0 \text{ or } Ax = \lambda x$$

Every matrix has n -eigenvalues (with multiplicities)

one term is $\prod_{i=1}^n (a_{ii} - \lambda)$. Polynomial of degree n in λ

Def: $\lambda_1(A), \lambda_2(A), \dots, \lambda_n(A)$

$$\lambda_1(A) + \lambda_2(A) + \dots + \lambda_n(A) = \text{trace}(A)$$

Examples:

$$\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix} = A \quad \det(A - \lambda I) = (1 - \lambda)^2 \quad \lambda = 1$$

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Rightarrow \begin{pmatrix} s \\ 0 \end{pmatrix} \quad s \text{ is any no.}$$

When does A have n -linear indep eigenvectors?

1. If eigenvalues are all distinct
2. if $A = A^T$ matrix has n linear indep eigenvectors.
3. If $A A^T = A^T A$

If A has n linear indep eigenvectors $\exists$ a non singular matrix X s.t. $A = X \Lambda X^{-1}$

where Λ is diagonal

$A X = X \Lambda$ this says that the column of X are eigenvectors

$$\Lambda = \begin{pmatrix} \lambda_1 & & 0 \\ & \lambda_2 & \\ 0 & & \ddots \\ & & & \lambda_n \end{pmatrix}$$

Theorem: Let $x^{(0)}, x^{(1)}, \dots$ be a sequence of column vectors generated by iteration scheme

$$x^{(m+1)} = B x^{(m)} + c$$

Then $x^{(m)} \rightarrow x$ where $x = Bx + c$ iff $\rho(B) < 1$

Proof: let

$$e^{(m)} = x^{(m)} - x$$

$$e^{(m+1)} = B e^{(m)} = B^2 e^{(m-1)} = B^{m+1} e^{(0)}$$

When B has n linear indep eigenvectors then $B^{m+1} e^{(0)} \rightarrow 0$

if $\rho(B) < 1 \quad m \rightarrow \infty$

if B has n linear indep eigenvectors

$$e^{(0)} = \sum_{j=1}^n \alpha_j \xi_j \quad A \xi_j = \lambda_j \xi_j$$

$$B e^{(0)} = \sum_{j=1}^n \alpha_j B \xi_j = \sum_{j=1}^n \alpha_j \lambda_j \xi_j$$

$$B^{(m)} e^{(0)} = \sum_{j=1}^n \alpha_j \lambda_j^{(m)} \xi_j \rightarrow 0 \quad \text{if } \rho(B) < 1$$

because $\lambda_i^{(m)} \rightarrow 0 \quad m \rightarrow \infty \quad i=1, \dots, n$

How is $\rho(A)$ related to norm A

where $\| \cdot \|$ is some natural norm. i.e. a matrix norm derived or induced from a vector norm by

$$\|A\| = \max_{x \neq 0} \frac{\|Ax\|}{\|x\|}$$

Result: $\rho(A) \leq \|A\|$

Proof:

Let $|\lambda| = \rho(A) \quad Au = \lambda u$

$$\|A\| = \max_{x \neq 0} \frac{\|Ax\|}{\|x\|} \geq \frac{\|Au\|}{\|u\|} = \frac{|\lambda| \|u\|}{\|u\|} = \rho(A)$$

Result: Given any $\epsilon > 0$ $\exists$ a natural matrix norm

$\rho(A) \leq \|A\| \leq \rho(A) + \epsilon$

Examples: $\begin{pmatrix} .9 & 1000 \\ 0 & .9 \end{pmatrix} \quad \|A\|_{\infty} = 1000.9$
 $\|A\|_1 = 1000.9$

$\rho(A) = .9$

If $A = A^T$ then $\rho(A) = \|A\|_2$

In general $\|A\|_2 = \sqrt{\rho(AA^T)}$

if $A = A^T \quad \rho(AA^T) = \rho(A^2) = \rho^2(A)$

if $Ax = \lambda x, \quad A^2 x = A(Ax) = A(\lambda x) = \lambda^2 x$

Proof of $\|A\|_2 = \sqrt{\rho(AA^T)}$

Put $B = AA^T$ then $B^T = B$

Recall B has n -lin indep. eigenvectors

$$Bu_s = \lambda_s u_s \quad s=1, \dots, n$$

Since B is sym $u_s^T u_{s'} = \delta_{ss'} = \begin{cases} 1 & s=s' \\ 0 & s \neq s' \end{cases}$

$$\|A\|_2^2 = \max_{x \neq 0} \frac{\|Ax\|_2^2}{\|x\|_2^2} = \max_{\|x\|_2=1} \|Ax\|_2^2 = \max_{\|x\|_2=1} (Ax)^T \cdot (Ax)$$

$$= \max_{\|x\|_2=1} x^T (A^T A) x = \max_{\|x\|_2=1} x^T B x = \rho(AA^T)$$

Let λ be the eigenvalue of $A^T A = B$ for which $|\lambda| = \rho(B)$ & u

the corresp. eigenvector. Then

Arrange $\|u\|_2 = 1$

$$\max_{\|x\|_2=1} x^T (A^T A) x \geq u^T B u = u^T \lambda u = \lambda u^T u = \lambda = \rho(B)$$

To get $\leq$ use the

representation $x = \sum_{j=1}^n x_j u_j$ plug in & it works

Example:

$$A = \begin{pmatrix} 1 & 1 \\ 0 & -1 \end{pmatrix}$$

$$\|A\|_\infty = 2 \quad \|A\|_1 = 2$$

$$A^T = \begin{pmatrix} 1 & 0 \\ 1 & -1 \end{pmatrix}$$

$$\|A\|_2 = \sqrt{\frac{3+\sqrt{5}}{2}}$$

* Check get eigenval of AA^T

$$AA^T = \begin{pmatrix} 2 & -1 \\ -1 & 1 \end{pmatrix}$$

$$\begin{vmatrix} 2-\lambda & -1 \\ -1 & 1-\lambda \end{vmatrix} = (2-\lambda)(1-\lambda) - 1 = 1 - 3\lambda + \lambda^2 = \frac{3+\sqrt{5}}{2} \text{ etc.}$$

$$* A = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$$

$$\|A\|_1 = 3 \quad \|A\|_2 = \rho(A)$$

$$A^T = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}$$

$$\|A\|_\infty = 3 = 3$$

③ Relaxation iteration scheme

HW P.180 3.91-95 due 2 wks

$$Ax=b$$

$$Ax-b=0$$

$$x = x - C^{-1}(Ax-b) \quad C \text{ is invertible}$$

$$x = (I - C^{-1}A)x + C^{-1}b$$

Fixed Pt form $x = Bx + c$

Iterates $x^{(m+1)} = Bx^{(m)} + c$ $x^{(0)}$ is given

$\rho(B)$ is what controls convergence of scheme.

a) Jacobi corresponds $C = \hat{D}$

$$x^{(m+1)} = (I - \hat{D}^{-1}A)x^{(m)} + \hat{D}^{-1}b$$

$$x_i^{(m+1)} = \sum_{j \neq i} -\frac{a_{ij}}{a_{ii}} x_j^{(m)} + \frac{b_i}{a_{ii}}$$

Convergence: if $\|I - \hat{D}^{-1}A\|_{\infty} < 1$ $\rho(I - \hat{D}^{-1}A)$

$\| \cdot \|_1$ A is strictly column diag dominant

$\| \cdot \|_{\infty}$ A " " row " "

b) Gauss-Seidel $C = \hat{L} + \hat{D}$

$$x_i^{(m+1)} = \frac{1}{a_{ii}} \left[\sum_{j=1}^{i-1} a_{ij} x_j^{(m+1)} + \sum_{j=i+1}^n a_{ij} x_j^{(m)} - b_i \right]$$

Convergence: If $\kappa = \max_{1 \leq i \leq n} \sum_{j \neq i} \left| \frac{a_{ij}}{a_{ii}} \right| < 1$

A is strictly row diagonally dominant.

c) Iterative Correction method

$C = \tilde{A}^{-1}$ = Approx inverse from Gauss elim.

$$x^{(m+1)} = (I - \tilde{A}^{-1}A) x^{(m)} + \tilde{A}^{-1}b$$

$x^{(0)}$ arbitrary

Let $x^{(0)}$ be arbitrary define

$$r^{(0)} = b - Ax^{(0)} \quad \text{residual}$$

Solve by GE

$$Ae^{(0)} = r^{(0)}$$

$$x^{(1)} = x^{(0)} + e^{(0)}$$

$$r^{(1)} = b - Ax^{(1)}$$

$$Ae^{(1)} = r^{(1)}$$

$$x^{(2)} = x^{(1)} + e^{(1)}$$

$$e^{(0)} = \tilde{A}^{-1} r^{(0)}$$

$$e^{(1)} = \tilde{A}^{-1} r^{(1)}$$

in principle this is faster & more accurate than the others.

4) Power Method for computing $\rho(A)$

where $|\lambda_1| > |\lambda_2| \geq |\lambda_3| \geq \dots \geq |\lambda_n| \quad n \leq m$

$$\rho(A) = |\lambda_1|$$

Method: Let z be a vector, ^{to be described} compute the sequence of vectors $z, Bz, B^2z, B^3z, \dots, B^mz$

Let $y_1, \dots, y_n$ be n lin indep. eigenvectors corresp $\lambda_1, \lambda_2, \dots, \lambda_n$

$$\text{Let } z = \sum_{i=1}^n y_i \quad B^m z = \sum_{i=1}^n \lambda_i^m y_i$$

$$\left(\frac{1}{\lambda_1} B\right)^m z = y_1 + \sum_{i=2}^n \left(\frac{\lambda_i}{\lambda_1}\right)^m y_i \quad \text{as } m \rightarrow \infty \quad \rho_{\text{lim}} \rightarrow 0$$

$$\left(\frac{1}{\lambda_1} B\right)^m z \sim y_1$$

$$z^{(m)} = \frac{B^m z}{i^{th} \text{ entry of } B^m z} = \left(\frac{1}{\lambda_1} B \right)^m z \sim \frac{y_1}{i^{th} \text{ entry of } y_1} = y_1^* \quad \text{provided } \neq 0$$

$$B z^{(m)} \sim B y_1^* = \lambda_1 y_1^* \quad \lambda_1 \approx \frac{i^{th} \text{ entry of } B z^{(m)}}{i^{th} \text{ entry of } B^m z}$$

$$\begin{bmatrix} 1 & 2 & 0 \\ 2 & 1 & 0 \\ 0 & 0 & -1 \end{bmatrix} \begin{bmatrix} -\lambda & & \\ 0 & -\lambda & \\ 0 & & -\lambda \end{bmatrix} \begin{bmatrix} 1-\lambda & 2 & 0 \\ 2 & 1-\lambda & 0 \\ 0 & 0 & -1-\lambda \end{bmatrix} = -(\lambda+1)[(1-\lambda)^2 - 4] = 0 \quad \lambda = -1$$

$$\lambda^2 - 2\lambda + 3 \quad (\lambda-3)(\lambda+1)$$

$$\lambda_1 = 3 \quad \lambda_2 = -1 \quad \lambda_3 = -1$$

$$\begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \quad \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} \quad \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

no. of iter	0	1	2	3	4	5	
$B^m z$	$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$	$\begin{pmatrix} 5 \\ 4 \\ -3 \end{pmatrix}$	$\begin{pmatrix} 13 \\ 14 \\ 3 \end{pmatrix}$	$\begin{pmatrix} 41 \\ 40 \\ 3 \end{pmatrix}$	$\begin{pmatrix} 121 \\ 122 \\ 3 \end{pmatrix}$	$\begin{pmatrix} 365 \\ 365 \\ 3 \end{pmatrix}$	
$z = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$	m	1	2	3	4	5	
	λ_1	5	$13/5$	$41/13$	$121/41$	$365/121$	$i=1$
	λ_2		$14/4$	$40/14$	$122/40$	$365/122$	$i=2$

Nov 27, 1972

Given $n+1$ pts: $x_0, x_1, \dots, x_n$ & the values of $f(x)$ at these pts find a polynomial of degree $\leq n$. $\exists$.

$$p(x_i) = f(x_i) \quad i=0, \dots, n$$

Solution (Lagrange)

Define

$$l_k(x) = \prod_{i=0}^n \frac{(x-x_i)}{(x_k-x_i)} = \text{poly of degree } \leq n$$

$$l_k(x_j) = \begin{cases} 0 & k \neq j \\ 1 & k = j \end{cases} = \delta_{kj}$$

Solution $p_n(x) = \sum_{k=0}^n f(x_k) l_k(x)$

$$p_n(x_j) = \sum_{i=0}^n f(x_i) l_i(x_j) = f(x_j), \therefore p_n(x) \text{ is a solution}$$

& it is unique

Assume $q_n(x)$ is another; take the difference $p_n(x) - q_n(x) = h_n(x)$ is a poly of degree n with $n+1$ zeroes is impossible $\therefore \equiv 0$

Example: 2 point $n=1$ x_0, x_1

$$l_0(x) = \frac{x-x_1}{x_0-x_1}, \quad l_1(x) = \frac{x-x_0}{x_1-x_0}$$

$$p_2(x) = f(x_0) \frac{x-x_1}{x_0-x_1} + f(x_1) \frac{x-x_0}{x_1-x_0}$$

$$p_2(x) = f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x - x_0)$$

$p_2(x)$ is the line through $(x_0, f(x_0))$ & $(x_1, f(x_1))$

$n=2$

$$p_3(x) = \frac{f(x_0)(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} + \frac{f(x_1)(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)}$$

$$+ \frac{f(x_2)(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)}$$

$$= \frac{1}{x_0-x_2} \left[f(x_0) \frac{(x-x_1)(x-x_2)}{x_0-x_1} + f(x_1) \frac{(x-x_0)(x-x_2)(x_0-x_2)}{(x_1-x_0)(x_1-x_2)} \right]$$

$$+ f(x_2) \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)}$$

$$f(x_0) + \frac{f(x_1) - f(x_0)}{x_1 - x_0} (x - x_0) + \left[\frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0} \right] \frac{(x - x_0)(x - x_1)}{x_2 - x_0}$$

In general let $f[x_0] = f(x_0)$

$$\& f[x_0, x_1, \dots, x_n] = \frac{f[x_1, x_2, \dots, x_n] - f[x_0, x_1, \dots, x_{n-1}]}{x_n - x_0}$$

$$f[x_0, x_1] = \frac{f[x_1] - f[x_0]}{x_1 - x_0}$$

$$f[x_0, x_1, x_2] = \frac{f[x_0, x_2] - f[x_0, x_1]}{x_2 - x_0}$$

We expect then

$$p_n(x) = \sum_{k=0}^n f(x_k) l_k(x) \quad \text{Newton's form}$$

$$= f[x_0] + f[x_0, x_1](x - x_0) + f[x_0, x_1, x_2](x - x_0)(x - x_1) + \dots + f[x_0, x_1, \dots, x_n](x - x_0)(x - x_1) \dots (x - x_{n-1})$$

Where $l_k(x)$ & $f[\]$ have been defined

$$\oplus P_{n+1} = P_n + f[x_0, \dots, x_n, x_{n+1}](x - x_0)(x - x_1) \dots (x - x_n)$$

$$\text{Let } \psi(x) = (x - x_0)(x - x_1) \dots (x - x_n)$$

then

$$l_k(x) = \frac{\psi(x)}{(x - x_k) \psi'(x_k)} = \prod_{\substack{j=0 \\ j \neq k}}^n \frac{(x - x_j)}{(x_k - x_j)}$$

$$\therefore p_n(x) = \psi(x) \sum_{k=0}^n \frac{f(x_k)}{(x - x_k) \psi'(x_k)}$$

If $f(x) \equiv 1$

$$P_n(x) = \psi(x) \sum_{k=0}^n \frac{1}{(x - x_k) \psi'(x_k)} = 1 = \sum_{k=0}^n l_k(x)$$

Lemma

if $f(x) = p_n^*(x)$ a poly of degree $\leq n$ and $p_n(x) = \text{Lagrange Int poly with } f = p_n^*$

Then $p_n(x) \equiv p_n^*(x)$

Proof

$p_n(x) - p_n^*(x) = h(x) = \text{poly of degree } \leq n$ & vanishes at $n+1$ pts & therefore must be ident $\equiv 0$.

To get $1 = \sum l_k(x)$ apply lemma with $n=0$

Observe further $f[x_0, x_1, \dots, x_n] = \sum_{k=0}^n \frac{f(x_k)}{\prod_{j \neq k} (x_k - x_j)}$

$\therefore f[\]$ is a symmetric function of $(x_0, x_1, \dots, x_n)$

To show this fact use $\otimes$ $p_n = p_{n-1} + f[\]() () \dots$
& equate coeff of x^n of both sides.
 $= f[x_0, \dots, n] x^n$

$$l_k(x) = \frac{1}{\prod_{j \neq k} (x_k - x_j)} x^n + \dots$$

$$\sum_{k=0}^n \frac{f(x_k)}{\prod_{j \neq k} (x_k - x_j)} x^n + \dots$$

Nested Multiplication

$$p(x) = a_0 + a_1 x + \dots + a_n x^n$$

set $b_n = a_n$

for $k = n-1, n-2, \dots, 0$ Do

$$b_k = a_k + (-z) b_{k+1}$$

$$\text{Then } p(x) = b_0 + (x-z) \underbrace{(b_1 + b_2 x + \dots + b_n x^{n-1})}_{q(x)}$$

$$p(z) = b_0$$

Generalize to Newton type poly

$$p(x) = a_0 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + \dots + a_n(x-x_0)\dots(x-x_{n-1})$$

Find $p(z)$

$$\text{Set } b_n = a_n$$

For $k = n-1, n-2, \dots, 0$ DO

$$b_k = a_k + (z - x_k) b_{k+1}$$

$$\text{Then: } p(x) = b_0 + (x-z)(b_1 + (x-x_0)(b_2 + (x-x_0)(x-x_1)(b_3 + \dots + (x-x_{n-1})b_n))$$

$$\therefore p(z) = b_0$$

To prove nested multiplication algorithm

$$\begin{aligned} p(x) &= a_0 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + \dots + a_n(x-x_0)\dots(x-x_{n-1}) \\ &= (b_0 - (z-x_0)b_1) + [(b_1 - (z-x_1)b_2)(x-x_0) + (b_2 - (z-x_2)b_3)](x-x_0)(x-x_1) \\ &\quad b_0 + b_1(x-z) + b_2(x-x_0)(x-z) + \dots \quad \text{QED} \end{aligned}$$

Note & recall the Newton form: $p(x) = \sum a_i(x-x_i)(x-x_{i+1})\dots$

$$\therefore a_i = f[x_0, x_1, x_2, \dots, x_i]$$

$$b_n = f[x_0, x_1, \dots, x_n]$$

$$f[x_0, x_1, \dots, x_k] = b_k - (z - x_k) b_{k+1} \quad \text{set } k = n-1$$

$$f[x_0, x_1, \dots, x_{n-1}] = b_{n-1} - (z - x_{n-1}) f[x_0, x_1, \dots, x_n]$$

$$\therefore f[x_0, x_1, \dots, x_n] = \frac{b_{n-1} - f[x_0, x_1, \dots, x_{n-1}]}{z - x_{n-1}}$$

Suppose $f[x_0, \dots, x_n]$ was not given as a formula

$$\text{Let } z = x_n$$

ok above

$$f[x_0, x_1, \dots, x_n] = b_{n-1} - f[x_0, x_1, \dots, x_{n-1}]$$

$$x_n = x_{n-1}$$

$$p_n(x) = \sum_{k=0}^n f[x_0, x_1, \dots, x_k] \prod_{j=0}^{k-1} (x - x_j)$$

Problem: Given $x_0, \dots, x_n$ & $f(x_0), \dots, f(x_n)$
 Find an efficient method for computing diff.

$$\begin{array}{ccccccc} x_0 & f[x_0] & & & & & \\ x_1 & f[x_1] & f[x_0, x_1] & & & & \\ & & f[x_1, x_2] & f[x_0, x_1, x_2] & & & \\ & & & & & & \\ & & & & & & \\ x_n & f[x_n] & f[x_{n-1}, x_n] & f[x_{n-2}, x_{n-1}, x_n] & \dots & f[x_0, \dots, x_n] \end{array}$$

write

$$p_n(x) = \sum_{k=0}^n f[x_k, x_{k+1}, \dots, x_n] \prod_{j=k+1}^n (x - x_j)$$

Algorithm for computation of coeff in Newton

Given $x_0, x_1, \dots, x_n$; $f(x_0), \dots, f(x_n)$

Initialize: $d_i = f(x_i)$ $i = 0, \dots, x_n$

For $k = 1, 2, \dots, n$ Do
 for $i = 0, 1, \dots, n-k$ Do
 $d_i = (d_{i+1} - d_i) / (x_{i+k} - x_i)$

Then $d_k = f[x_k, x_{k+1}, \dots, x_n]$, $i = 0, 1, \dots, n$

for $k=1$ $i=0, \dots, n-1$

$$d_0 = \frac{f(x_1) - f(x_0)}{x_1 - x_0} \quad \dots \quad d_{n-1} = \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$$

for $k=2$ $i=0, \dots, n-2$

$$d_{n-2} = f[x_{n-2}, x_{n-1}, x_n]$$

Example

$$\frac{1}{1+x^2}$$

$$-5 \leq x \leq 5$$

Use poly interp

$$x_0, x_1, \dots, x_n$$

$$x_k = -5 + \frac{10k}{n} \quad n \text{ even}$$

let E_n be max $|f(x) - P_n(x)|$
 $-5 \leq x \leq 5$

n	E_n (exp)
2	6.4 (-1)
4	4.3 (-1)
6	6.1 (+1)
8	1 (-2)
10	1
12	1
14	1
16	1.4 (1)

$$P_2(x) = \frac{1}{26} + \frac{5}{26}(x+5) - \frac{1}{26}(x-5)x$$

use low degree poly in smaller intervals.

Error: $x_0, x_1, \dots, x_n \in [a, b]$

$$f(x_0), \dots, f(x_n)$$

$$f(x) - P_n(x) = e_n(x)$$

To get a formula for $e_n(x)$

let $\bar{x}$ be another interp pt $(x_0, \dots, x_n, \bar{x})$

$$P_{n+1}(x) = P_n(x) + f[x_0, \dots, x_n, \bar{x}](x-x_0)(x-x_1)\dots(x-x_n)$$

set $x = \bar{x}$

$$f(\bar{x}) = P_n(\bar{x}) + f[x_0, \dots, x_n, \bar{x}](\bar{x}-x_0)(\bar{x}-x_1)\dots(\bar{x}-x_n)$$

$$\therefore e_{n+1} = f[x_0, x_1, \dots, x_n, x](x-x_0)(x-x_1)\dots(x-x_n)$$

$$\exists \xi \quad e_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} (x-x_0)\dots(x-x_n)$$

$\min x_i < \xi < \max x_i$

Dec 4, 1972

$$\begin{bmatrix} 4 & -1 \\ -1 & 4 \end{bmatrix} \quad 3.95 \quad \text{program power method} \\ \text{and different starting vectors}$$

1. Review Interpol. Poly.
2. Error radius of convergence & lemniscate of convergence
3. Forward & Backward differences
4. Spline Interpolation (Cubic)
5. Least Squares & Orthogonal polynomials.

1. $x_0, x_1, \dots, x_n$

$f(x_0), f(x_1), \dots, f(x_n)$

Construct a poly $\leq n \Rightarrow p(x_i) = f(x_i)$

Lagrange

$$l_k(x) = \prod_{\substack{i=0 \\ i \neq k}}^n \frac{(x-x_i)}{(x_k-x_i)} \quad l_k(x_j) = \delta_{kj}$$

$$P_n(x) = \sum_{k=0}^n f(x_k) l_k(x)$$

Newton's form

$$f[x_0] = f(x_0) \quad f[x_0, \dots, x_k] = \frac{f[x_1, \dots, x_k] - f[x_0, \dots, x_{k-1}]}{x_k - x_0}$$

$$P_n(x) = \sum_{k=0}^n f[x_0, \dots, x_k] (x-x_0) \dots (x-x_{k-1})$$

$$P_{n+1}(x) = P_n(x) + f[x_0, x_1, \dots, x_{n+1}] (x-x_0) \dots (x-x_n)$$

2) Error: $e_n(x) = f(x) - P_n(x) = f[x_0, \dots, x_n, x] (x-x_0) (x-x_1) \dots (x-x_n)$

Construct $P_{n+1}(x) = P_n(x) + f[x_0, \dots, x_n, \bar{x}](x-x_0) \dots (x-x_n)$

$$P_{n+1}(\bar{x}) = f(\bar{x})$$

$$e_n(x) = f(x) - P_n(x) = f[x_0, \dots, x_n, x](x-x_0) \dots (x-x_n)$$

Theorem: $f[x_0, x_1, \dots, x_n] = f^{(n)}(\xi) \quad \min x_i \leq \xi \leq \max x_i$

For $n=1$ $f[x_0, x_1] = \frac{f(x_1) - f(x_0)}{x_1 - x_0} = f'(\xi) \quad \text{M.V.T.}$

For $n=2$ $\frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0} = \frac{f^{(1)}(\xi_2) - f^{(1)}(\xi_1)}{x_2 - x_0} = f^{(2)}(\xi_2)$

$e_n(x)$ has $n+1$ zeroes in $[a, b]$

$e_n'(x)$ has n zeroes in $[a, b]$ by Rolle's Theorem

$e_n''(x)$ has $n-1$ zeroes in $[a, b]$

$e_n^{(n)}(x)$ has 1 zero in $[a, b]$

$$\therefore e_n^{(n)}(\xi) = 0 \quad \xi \in [a, b]$$

$$e_n^{(n)}(\xi) = f^{(n)}(\xi) - \frac{P_n^{(n)}(x)}{n!} = 0 \quad \text{QED}$$

Error: $e_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \prod_{i=0}^n (x-x_i) \quad \min(x_i, x) \leq \xi \leq \max(x_i, x)$

Recall Taylor's expansion error

Find x_i 's so that $g(x; x_0, \dots, x_n)$ is max for all

$$g(x) = \prod_{i=0}^n (x-x_i)$$

$$\min_{x \in [a, b]} |g(x; x_0, \dots, x_n) - g(x; x_0^*, x_1^*, \dots, x_n^*)| = \min \text{ over all}$$

3. Forward & Backward Differences:

Interpolation pts equally spaced

$$x_k = x_0 + kh \quad k=0, 1, \dots, n$$

For a general $x = x_0 + sh$ s not an integer; $f(x_0 + sh) = f_s$

$$\Delta^0 f_s = f_s \quad \Delta^i f_s = \Delta^{i-1} f_{s+1} - \Delta^{i-1} f_s$$

$$P_n(x) = \sum_{k=0}^n f[x_0, x_1, \dots, x_k] \prod_{j=0}^{k-1} (x - x_j)$$

$$f[x_k, x_{k+1}, \dots, x_{k+n}] = \Delta^n f_k \left(\frac{1}{n! h^n} \right)$$

for $n=0$

$$f[x_k] = f(x_k) = f_k = f(x_0 + kh)$$

Assume true for n & induce for $n+1$

$$f[x_k, \dots, x_{k+n}, x_{k+n+1}] = f[x_{k+1}, x_{k+2}, \dots, x_{k+n}, x_{k+n+1}] - f[x_k, x_{k+1}, \dots, x_{k+n}]$$

$$= \frac{1}{n! h^n} \Delta^n f_{k+1} - \frac{1}{n! h^n} \Delta^n f_k$$

$$= \frac{1}{(n+1)! h^{n+1}} [\Delta^n f_{k+1} - \Delta^n f_k] = \Delta^{n+1} f_k \frac{1}{(n+1)! h^{n+1}}$$

Plug into Newton's form:

$$P_n(x_0 + sh) = \sum_{i=0}^n f[x_k, x_{k+1}, \dots, x_{k+i}] \prod_{j=0}^{i-1} (x - x_{k+j})$$

$$= \sum_{i=0}^n \frac{1}{i! h^i} \Delta^i f_k h^i \prod_{j=0}^{i-1} (s - k - j) = \sum_{i=0}^n \Delta^i f_k \prod_{j=0}^{i-1} (s - k - j)$$

$$= \binom{s-k}{i}; \quad \binom{y}{i} = \frac{y(y-1)\dots(y-i+1)}{1 \dots i}$$

$$\therefore p(x_0 + sh) = \sum_{i=0}^n \binom{s-k}{i} \Delta^i f_k$$

Example		Δf	$\Delta^2 f$	$\Delta^3 f$	$\Delta^4 f$
x	f(x)				
0	1				
1	5	4			
2	31	26	22		
3	121	90	64	42	
4	341	220	120	66	24

find $f(5)$ $n=4$ $s=5$ $k=0$

$$\binom{5}{0} = 1 \quad \binom{5}{1} = 5 \quad \binom{5}{2} = 10 \quad \binom{5}{3} = 10 \quad \binom{5}{4} = 5$$

$$\binom{5}{5} = 1$$

$$f(5) = 1 \cdot 1 + 4 \cdot 5 + 10 \cdot 22 + 10 \cdot 42 + 5 \cdot 24 = 1 + 20 + 220 + 420 + 120 = 781$$

Backward form of Newton Interpolation polyn.

$$p_n(x_0 + sh) = \sum_{i=0}^n f[x_{n-i}, x_{n-i+1}, \dots, x_n] \prod_{j=0}^{i-1} (x - x_{n-j})$$

$$= \sum_{i=0}^n \frac{1}{i! h^i} \Delta^i f_{n-i} \prod_{j=0}^{i-1} h(s - n + j)$$

$$= \sum_{i=0}^n \frac{1}{i!} \Delta^i f_{n-i} \prod_{j=0}^{i-1} (s - n + j)$$

$$= \sum_{i=0}^n \frac{(-1)^i}{i!} \Delta^i f_{n-i} \prod_{j=0}^{i-1} (n - s - j)$$

$$= \sum_{i=0}^n (-1)^i \Delta^i f_{n-i} \binom{n-s}{i}$$

Spline Interpolation (Cubic)

Piecewise polyn interpolation

$$x_1, x_2, \dots, x_{n+1}$$

$$a = f(x_1), \dots, f(x_{n+1}) = b$$

Problem: find $g_3(x)$ = cubic polyn interpolating the points whose $f(x)$ is defined at given pts.

Notation: in each interval $\{x_i, x_{i+1}\}$ $i=1, \dots, n$ construct

$$P_i(x) = c_{i,1} + c_{i,2}(x-x_i) + c_{i,3}(x-x_i)^2 + c_{i,4}(x-x_i)^3$$

To determine coeff:

$$P_i(x_i) = f(x_i) \quad P_i(x_{i+1}) = f(x_{i+1}) = P_{i+1}(x_i) \quad \therefore g_3(x) \text{ cont.}$$

Various possibilities to obtain the other two constants in P_i

1) Hermite interpolation (requires $f'(x_i)$)

$$\text{require } P_i'(x_i) = f'(x_i) \quad P_i'(x_{i+1}) = f'(x_{i+1})$$

$$P_i(x) = f(x_i) + f[x_i, x_i](x-x_i) + f[x_i, x_i, x_{i+1}](x-x_i)^2 + \underbrace{f[x_i, x_i, x_{i+1}, x_{i+1}]}_{(x-x_i) = \Delta x_i} (x-x_i)^2 (x-x_{i+1})$$

$$c_{i,1} = f(x_i)$$

$$c_{i,2} = f'(x_i) = S_i$$

$$\begin{aligned} c_{i,3} &= f[x_i, x_i, x_{i+1}] = f[x_i, x_i, x_{i+1}, x_{i+1}] \Delta x_i \\ &= \frac{f[x_i, x_{i+1}] - f[x_i, x_i]}{x_{i+1} - x_i} - c_{i,4} \Delta x_i = \frac{f[x_i, x_{i+1}] - S_i}{\Delta x_i} - c_{i,4} \Delta x_i \end{aligned}$$

$$c_{i,4} = f[x_i, x_i, x_{i+1}, x_{i+1}] = \frac{S_{i+1} + S_i - 2f[x_i, x_{i+1}]}{(\Delta x_i)^2}$$

$g_3(x)$ is now determined $\Rightarrow g_3(x) \in C^1$ in $[a, b]$

2) Bessel interpol $S_i = f[x_{i-1}, x_i]$ @ end pts extra data

3) Spline Interpolation: Fix $C_{i,1,2,3,4} \Rightarrow g_3(x)$ has 2 derivatives

Now $P_i(x)$ are determined globally

Require $P_{i-1}''(x_i) = P_i''(x_i)$

$$2C_{3,i-1} + 6C_{4,i-1} \Delta x_{i-1} = 2C_{3,i} \quad i=2, \dots, N$$

put into formulas to get S_i

$$\frac{2(f[x_{i-1}, x_i] - S_{i-1})}{\Delta x_{i-1}} + 4C_{i-1,4} \Delta x_{i-1} = \frac{2(f[x_i, x_{i+1}] - S_i)}{\Delta x_i} - 2C_{i,4} \Delta x_i$$

$$(\Delta x_i) S_{i-1} + 2(\Delta x_{i-1} + \Delta x_i) S_i + (\Delta x_{i-1}) S_{i+1} = 3(f[x_{i-1}, x_i] \Delta x_i + f[x_i, x_{i+1}] \Delta x_{i-1})$$

$i=2, \dots, N$

$N-1$ eqs $N+1$ S 's 2 must be given for b.c.

Then is finding system of LE for S 's.

Dec 11, 1972

Least Squares

Given $x_i \quad i=1, N \quad f(x_i) \quad i=1, N \quad \in [a, b]$

Choose $\varphi_1(x), \varphi_2(x), \dots, \varphi_k(x)$ Try to fit data

by expressions of the form

$$\sum_{i=1}^k c_i \varphi_i(x) \quad \text{choose } N \gg k$$

Fit in least square norm

$$e = \sum_{j=1}^N \sum_{i=1}^k (c_i \varphi_i(x_j) - f_j)^2$$

To find c 's diff & set deriv = 0

$$-2 \sum_{n=1}^N \left[f_n - \sum_{i=1}^N c_i \phi_i(x_n) \right] \phi_i(x_n) = 0$$

$$\sum_{i=1}^N c_i \left(\sum_{n=1}^N \phi_i(x_n) \phi_j(x_n) \right) = \sum_{n=1}^N f_n \phi_j(x_n)$$

Define

$$\langle f, g \rangle = \sum_{n=1}^N f_n g_n$$

Then $\sum_{n=1}^N \phi_i(x_n) \phi_j(x_n) = \langle \phi_i, \phi_j \rangle \therefore$ eqs for c 's take form

$$\sum_{i=1}^N \langle \phi_j, \phi_i \rangle c_i = \langle f, \phi_j \rangle \quad \text{normal eqs for } c\text{'s.}$$

Way to get c 's is to find ϕ 's so that they are orthogonal

$\Rightarrow \langle \phi_i, \phi_j \rangle = 0 \quad i \neq j$
 normalized $\langle \phi_i, \phi_i \rangle = 1 \quad i = j$

Then the c 's are given by

$$c_j = \frac{\langle f, \phi_j \rangle}{\langle \phi_j, \phi_j \rangle}$$

Problem now is to find the orthogonal fns $\phi_j(x)$.

Orthogonal Polynomials $[a, b]$, $f(x)$, $g(x)$ real fns on $[a, b]$

Inner Product

(i) $\int_a^b f g w \, dx = \langle f, g \rangle \quad w(x) > 0 \text{ (given) continuous}$

(ii) $\sum_{i=1}^N f g w = \langle f, g \rangle \quad \text{discrete}$

Examples

$$\langle f, g \rangle = \int_{-1}^1 f(x) g(x) dx \quad \langle f, g \rangle = \int_{-\infty}^{\infty} f(x) g(x) e^{-x^2} dx$$

$$\langle f, g \rangle = \int_{-1}^1 \frac{f(x) g(x)}{\sqrt{1-x^2}} dx \quad \langle f, g \rangle = \sum_{j=0}^M f(-1+jh) g(-1+jh) \quad h = \frac{2}{N}$$

f, g are orthogonal if $\langle f, g \rangle = 0$

Define $P_0(x), P_1(x), P_2(x), \dots$ for a system of orthog polyn
if (i) $P_k(x)$ is of degree $k = \alpha_k x^k + \text{poly deg} < k, \alpha_k \neq 0$
(ii) $\langle P_k, P_j \rangle = 0 \quad j \neq k$

Example: if $\langle f, g \rangle = \int_{-1}^1 f(x) g(x) dx$

then $P_0 = 1, P_1(x) = x, P_2(x) = 3x^2 - 1$ are orthog.

$$\langle P_0, P_1 \rangle = 0$$

$$\langle P_0, P_2 \rangle = 0$$

$$\langle P_1, P_2 \rangle = 0$$

To construct P 's need following general properties

1) If $p(x)$ is a poly of deg k then we can write

$$p(x) = \sum_{j=0}^k d_j \varphi_j(x) \quad \text{where } d_j = \langle p, P_j \rangle / \langle P_j, P_j \rangle$$

Example: $x^2 = \frac{1}{3} P_2(x) + \frac{1}{3} P_0(x)$

2) If $p(x)$ has degree $< k$ then $\langle p, P_k \rangle = 0$

$$p = \sum_{i=0}^{k-1} c_i \varphi_i \quad \therefore \langle p, P_k \rangle = \sum c_i \langle P_k, \varphi_i \rangle = c_i \langle P_k, P_i \rangle = 0 \quad \text{since } i \neq k$$

Aside $[a, b] = [0, 1]$

$$\varphi_j(x) = x^j \quad j = 0, 1, 2, \dots$$

$$\langle f, g \rangle = \int_a^b f g dx$$

$$H_n = \langle \phi_i, \phi_j \rangle = \int_0^1 x^{i+j} dx = \frac{1}{i+j+1} \quad i, j = 0, 1, \dots$$

Hilbert Ill-Conditioned Matrix.

(iii) $P_K(x)$ has exactly K real & distinct roots in $[a, b]$

Proof:

where $P_K(x) = \alpha_K (x - \xi_{1,K})(x - \xi_{2,K})(x - \xi_{3,K}) \dots (x - \xi_{K,K})$

$$a \leq \xi_{1,K} < \xi_{2,K} < \dots < \xi_{K,K} \leq b \quad \text{in } [a, b]$$

Suppose $\xi_{1,K}, \xi_{2,K}, \dots, \xi_{K,K}$ are zeros of $P_K(x)$ $K < K$

Consider $p_K(x)$ and the polyn $q = P_K(\xi) (x - \xi_{1,K})(x - \xi_{2,K}) \dots (x - \xi_{K,K})$
 have the same sign in $[a, b]$ where ξ is the largest of $(b, \xi_{K,K})$

degree $q(x) = r < K \therefore$ by (iii) $\langle q, P_K \rangle = 0$

$\therefore$ This is impossible since q & $P_K(x)$ has the same sign in $[a, b]$

& hence $\langle q, P_K \rangle = \int_a^b q P_K(x) w(x) dx > 0$

unless $r = 0$

(iv) Any 3 consecutive orthogonal polynomials satisfy the relation

$$P_{n+1}(x) = A_n (x - B_n) P_n(x) - C_n P_{n-1}(x) \quad n = 1, 2, \dots$$

$$P_0 \equiv 0$$

where

$$A_n = \frac{\alpha_{n+1}}{\alpha_n}$$

$$B_n = A_n \frac{\langle P_n, x P_n \rangle}{\langle P_{n-1}, P_{n-1} \rangle}; \quad C_n = A_n \frac{\langle P_n, P_n \rangle}{A_{n-1} \langle P_{n-1}, P_{n-1} \rangle}$$

Proof:

$$P_{n+1}(x) - \frac{\alpha_{n+1}}{\alpha_n} x P_n(x) = q_n(x) = \text{poly of degree } \leq n$$

$$\therefore q_n = \sum_{j=0}^n \gamma_j P_j(x) \quad \text{by (ii)}$$

$$\langle q_n, P_k \rangle = 0 \quad k \neq n-1, n$$

$$\langle P_{n+1} - \frac{\alpha_{n+1}}{\alpha_n} x P_n, P_k \rangle = \langle P_{n+1}, P_k \rangle - \frac{\alpha_{n+1}}{\alpha_n} \langle x P_n, P_k \rangle$$

if $k = n-1$ $x P_n$ poly of degree n $\therefore \langle P_n, x P_n \rangle \neq 0$ poly of $n+1$

$$\text{when } k = n \quad \langle q_n, P_n \rangle = \gamma_n \langle P_n, P_n \rangle = - \frac{\alpha_{n+1}}{\alpha_n} \langle P_n, x P_n \rangle$$

$$\text{when } k = n-1 \quad \langle q_n, P_{n-1} \rangle = \gamma_{n-1} \langle P_{n-1}, P_{n-1} \rangle = - \frac{\alpha_{n+1}}{\alpha_n} \langle P_n, x P_{n-1} \rangle$$

$$x P_{n-1} = \frac{\alpha_{n-1}}{\alpha_n} P_n + q_{n-1}$$

$$\therefore \gamma_{n-1} \langle P_{n-1}, P_{n-1} \rangle = - \frac{\alpha_{n+1}}{\alpha_n} \frac{\alpha_{n-1}}{\alpha_n} \langle P_n, P_n \rangle$$

$$\text{Set } A_n = \frac{\alpha_{n+1}}{\alpha_n} ; \quad \gamma_n = -B_n = A_n \frac{\langle P_n, x P_n \rangle}{\langle P_n, P_n \rangle}$$

$$\gamma_{n-1} = -C_n = \frac{A_n \langle P_n, P_n \rangle}{A_{n-1} \langle P_{n-1}, P_{n-1} \rangle}$$

Examples

$$1) \quad \langle f, g \rangle = \int_{-1}^1 f(x) g(x) dx ; \quad P_0(x) = 1 \quad P_k(+1) = 1$$

then $P_0, P_1, P_2, \dots$ are the Legendre Polyn

$$P_{n+1} = \frac{(2n+1)x P_n - n P_{n-1}}{n+1} \quad P_0 = 1 \quad P_{-1} = 0$$

2) Chebyshev Poly

$$\langle f, g \rangle = \int_{-1}^1 \frac{fg}{\sqrt{1-x^2}} dx \quad T_0(x) = 1 \quad T_{-1} = 0$$

Then $T_{n+1} = 2xT_n - T_{n-1}$

$$T_n(\cos \theta) = \cos n\theta$$

Nested Mult. Algorithm for evaluating $p(x)$
where

$$p(x) = d_0 P_0(x) + d_1 P_1(x) + \dots + d_k P_k(x)$$

($d_0, d_1, \dots, d_k$ given)

To find $p(x)$

$$p(x) = d_0 P_0 + d_1 P_1 + \dots + (d_{k-2} + C_{k-2} d_k) P_{k-2} + [d_{k-1} + d_k A_{k-1} (x - B_{k-1})] P_{k-1}$$

set $\bar{d}_k = d_k$

$$\bar{d}_{k-1} = d_{k-1} + \bar{d}_k A_{k-1} (x - B_{k-1})$$

$$p(x) = d_0 P_0 + \dots + (d_{k-2} + C_{k-2} \bar{d}_k) P_{k-2} + \bar{d}_{k-1} P_{k-1}$$

Repeat the above

Algorithm Nested Mult.:

Given $A_j, B_j, C_j \quad j=0, 1, \dots$, $d_0, d_1, \dots, d_k$

$\alpha_0 = P_0(1)$ and a pt x then

set $\bar{d}_k = d_k$

If $k=0$ stop

set $\bar{d}_{k-1} = d_{k-1} + \bar{d}_k A_{k-1} (x - B_{k-1})$

If $k=1$ stop

for $j = k-2, \dots, 0$ Do

$$\bar{d}_j = d_j + d_{j+1} A_j (\bar{x} - B_j) - d_{j+2} C_{j+1}$$

pr64

4.11-3

homework

Example: Back to Least Squares

Find polynomial $p^*(x)$ $\deg \leq 3 \Rightarrow \int_{-1}^1 [e^x - p^*(x)]^2 dx$
is min over all polyn of degree ≤ 3 .

i.e. find $p^*(x) = c_0^* + c_1^* x + c_2^* x^2 + c_3^* x^3$

$$\sum_{j=1}^n \langle \phi_i, \phi_j \rangle c_j = \langle f, \phi_i \rangle \quad f = e^x$$

$$\int_{-1}^1 f g dx = \langle f, g \rangle$$

Orthogonal Polyn are Legendre Polyn

$$\phi_j(x) \text{ and } p^*(x) = d_0^* P_0 + d_1^* P_1 + d_2^* P_2 + d_3^* P_3$$

$$\text{where } d_j^* = \frac{\langle e^x, P_j \rangle}{\langle P_j, P_j \rangle}$$

$$P_0 = 1 \quad P_1 = x \quad P_2 = \frac{3x^2 - 1}{2} \quad P_3 = \frac{5x^3}{2} (x^3 - \frac{3}{5}x)$$

$$\langle P_k, P_k \rangle = \frac{2}{2k+1} = \int_{-1}^1 P_k(x)^2 dx$$

$$\langle e^x, P_0 \rangle = \int_{-1}^1 e^x dx = e - \frac{1}{e}$$

$$\langle e^x, P_1 \rangle = \int_{-1}^1 e^x x dx = \frac{2}{e}$$

Generation & Use of discrete orthogonal polyn

$$x_1, x_2, \dots, x_n \in [a, b] \quad \langle g, h \rangle = \sum_{n=1}^N g(x_n) h(x_n) w(x_n)$$

Assume $A_n = 1$ $\alpha_0 = 1$

Step 0

$$P_0 = 1$$

$$\text{Compute } \langle P_0, P_0 \rangle = S_0 = \sum_{n=1}^N W(x_n)$$

If $N \geq 1$ & $W(x) > 0$ $S_0 \neq 0$

using $P_{n+1} = A_n(x - B_n)P_n - C_n P_{n-1}$; $A_n = 1$, $B_n = \frac{\langle x P_n, P_n \rangle}{\langle P_n, P_n \rangle}$

$$\begin{aligned} \text{Calculate } P_1(x) &= (x - B_0)P_0 = x - B_0 = x - \frac{\langle x P_0, P_0 \rangle}{S_0} \quad C_n = \frac{\langle P_n, P_n \rangle}{\langle P_{n-1}, P_{n-1} \rangle} \\ &= x - \frac{1}{S_0} \sum_{n=1}^N x_n W(x_n) \end{aligned}$$

Once $P_0, P_1, \dots, P_j$ calculated

Step j

$$\text{Compute } S_j = \langle P_j, P_j \rangle = \sum_{n=1}^N P_j(x_n)^2 W(x_n) \quad \text{if } > 0$$

$$B_j = \frac{\langle x P_j, P_j \rangle}{S_j} \quad (\neq 0 \text{ if there are more than } j \text{ distinct points among } x_1, x_2, \dots, x_n)$$

$$= \frac{1}{S_j} \sum P_j(x_n)^2 x_n W(x_n)$$

$$C_N = \frac{S_j}{S_{j-1}}$$

Continue.

Dec 18, 1977

Final Exam Mon Jan 22 11:30-1:30

Exam Closed Book - work up to Least Squares

Least Squares - orthogonal polyn.

Numerical Integration

Given $(x_1, f_1), (x_2, f_2), \dots, (x_n, f_n)$

and a sequence of fns $\phi_1(x), \phi_2(x), \phi_3(x), \dots$

Find $c_1, \dots, c_k$

$$\Rightarrow E(c_1, \dots, c_k) = \sum_{n=1}^N \left[f_n - \sum_{i=1}^k c_i \phi_i(x_n) \right]^2 = \min$$

$\frac{\partial E}{\partial c_j} = 0$ we obtain normal equations

$$\sum_{i=1}^k \langle \phi_i, \phi_j \rangle c_i = \langle f, \phi_j \rangle \quad \text{--- system of lin eqns}$$

$$\text{where } \langle \phi_j, \phi_i \rangle = \sum_{n=1}^N \phi_j(x_n) \phi_i(x_n) \quad \langle f, \phi_j \rangle = \sum_{n=1}^N \phi_j(x_n) f_n$$

Construct polynomials with property $\langle \phi_j, \phi_i \rangle = 0 \quad i \neq j$
polynomial class satisfies 4 properties & founded especially 4th
property of recurrence

$$\text{More general inner product } \langle f, g \rangle = \int_a^b f g w dx \quad w \geq 0$$

$$\langle f, g \rangle = \sum_{n=1}^N f_n g_n w(x_n)$$

$\langle, \rangle$ inner product has properties

$$\textcircled{1} \quad \langle f, f \rangle \geq 0 \quad = 0 \Rightarrow (\text{cont case } f=0)$$

$$\textcircled{2} \quad \langle a_1 f + a_2 f_2, g \rangle = a_1 \langle f, g \rangle + a_2 \langle f_2, g \rangle$$

$$\langle f, b_1 g_1 + b_2 g_2 \rangle = b_1 \langle f, g_1 \rangle + b_2 \langle f, g_2 \rangle$$

For a notion of $\langle, \rangle$ then $P_0(x), P_1(x), \dots$ is a seq of
orthogonal polynomial if (i) $P_j(x) = \alpha_j x^j + \text{polyn of degree } < j \quad \alpha_j \neq 0$
(ii) $\langle P_i, P_j \rangle = 0 \quad i \neq j$

Prop of orthog poly

- ① if $P(x)$ is poly of degree k then $p(x) = d_0 P_0(x) + d_1 P_1(x) + \dots + d_k P_k(x)$
- ② if $p(x)$ is poly $< k$ then $\langle p, P_k \rangle = 0$
- ③ $x \in [a, b]$ $P_k(x)$ has exactly k distinct zeros in $[a, b]$
- ④ Recurrence Relation

$$P_{i+1}(x) = A_i (x - B_i) P_i(x) - C_i P_{i-1}(x) \quad i=0,1,\dots$$

$$P_{-1}(x) \equiv 0 \quad A_i = \frac{x_{i+1}}{\alpha_i} \quad i=0,1,\dots \quad S_i = \langle P_i, P_i \rangle$$

$$B_i = \frac{1}{S_i} \langle x P_i, P_i \rangle \quad C_j = \frac{A_j}{A_{j-1}} \frac{S_j}{S_{j-1}} \quad j=1,\dots$$

Usually Assume $P_0(x) \equiv 1$

Possible normalizations:

Example Find a poly of degree ≤ 3 $\rightarrow \int_{-1}^1 [e^x - p(x)] dx = \min$

Inner product

$$\langle f, g \rangle = \int_{-1}^1 f g dx \quad P_0(x), P_1(x), P_2(x), \dots \text{ so called Legendre Poly}$$

$$p(x) = d_0 P_0 + d_1 P_1 + d_2 P_2 + d_3 P_3 \quad d_j = \frac{\langle e^x, P_j \rangle}{\langle P_j, P_j \rangle}$$

Construction of O.P. when $\langle f, g \rangle = \sum f_n g_n w(x_n) \quad w(x_n) > 0$

Assume $\alpha_i = 1$ and hence $A_i = 1 \quad P_{-1} = 0 \quad P_0 = 1$

$$P_{i+1} = (x - B_i) P_i - C_i P_{i-1}$$

$$S_i = \langle P_i, P_i \rangle, \quad B_i = \langle x P_i, P_i \rangle / S_i, \quad C_i = S_i / S_{i-1}$$

To construct $P_j(x)$

$$\textcircled{1} \text{ Set } P_0(x) = 1, \text{ compute } S_0 = \langle P_0, P_0 \rangle = \sum_{n=1}^N w(x_n)$$

$$B_0 = \frac{\langle x P_0, P_0 \rangle}{S_0} = \frac{1}{S_0} \sum_{n=1}^N x_n w(x_n)$$

② if $P_0, P_1, \dots, P_j$ are obtained

then compute:

$$S_j = \langle P_j, P_j \rangle = \sum_{n=1}^N P_j^2(x_n) W(x_n)$$

$$B_j = \frac{1}{S_j} \langle x P_j, P_j \rangle = \frac{1}{S_j} \sum_{n=1}^N x_n W(x_n) P_j^2(x_n)$$

Example:

$$x_n = 10 + \frac{(n-1)}{5} \quad n=1, \dots, 6$$

$$f_n = f(x_n) \quad f(x) = 10 - 2x + \frac{1}{10} x^2$$

Find by least squares a polyn of degree ≤ 2 fitting above data

$$p(x) = c_0 + c_1 x + c_2 x^2 \Rightarrow \sum [f_n - p(x_n)]^2 = \min$$

$$\text{Set } P_0(x) = 1 \quad S_0 = \sum_{n=1}^6 1 = 6 \quad B_0 = \frac{1}{6} \sum_{n=1}^6 10 + \frac{(n-1)}{5} = 10.5$$

$$P_1(x) = (x - 10.5)$$

$$S_1 = \langle P_1, P_1 \rangle = \sum_{n=1}^6 \left(\frac{n-1}{5} - 10.5 \right)^2 = .7$$

$$B_1 = \frac{1}{.7} \langle x P_1, P_1 \rangle = \frac{1}{.7} \sum_{n=1}^6 x (x - 10.5)^2 = 10.5$$

$$P_2(x) = (x - 10.5)^2 - \frac{.7}{6}$$

$$\text{One gets } d_j = \frac{\langle f, P_j \rangle}{\langle P_j, P_j \rangle}$$

$$d_0 = \frac{1}{6} \sum_{n=1}^6 \left[10 - 2 \left(10 + \frac{n-1}{5} \right) + \frac{1}{10} \left(10 + \frac{n-1}{5} \right)^2 \right] = 0.0366$$

Fortran Implementation for algorithm above

DATA $X(n) = x_n$, $F(n) = f_n$, $W(n) = w(x_n)$, $n=1, \dots, NPOINT$

$NTERMS-1$ = degree of polyn

To construct $B(J)$, $C(J)$ $J=1, \dots, NTERMS-1$

$D(J)$, $J=1, \dots, NTERMS$

$$p(x) = D(1) + D(2) P_1(x) + D(3) P_2(x) + \dots + D(NTERMS) P_{NTERMS-1}(x)$$

With other polyn eval routine the above gives complete answer to the above original problem.

$$\text{Error}(n) = [f_n - p_n^*]$$

SUBROUTINE GRTPOL (NP, X, F, W, ER, PJM1, PJ)

DIM X(100), F(100), W(100), ER(100), PJ(20), PJM1(20), S(20)

COMMON / POLY /, NT, B(20), C(20), D(20)

<pre> DO 9 J=1, NT B(J)=0. D(J)=0. 9 S(J)=0. C(1)=0. DO 10 I=1, NP B(1)=B(1)+X(I)*W(I) D(1)=D(1)+F(I)*W(I) 10 S(1)=S(1)+W(I) D(1)=D(1)/S(1) DO 11 I=1, NP ER(I)=F(I)-D(I) IF (NT.EQ.1) RETURN B(1)=B(1)/S(1) DO 12 I=1, NP PJM1(I)=1. 12 PJ(I)=X(I)-B(1) J=1 20 J=J+1 </pre>	<pre> DO 21 I=1, NP S(J)=S(J)+PJ(I)*PJ(I)*W(I) B(J)=B(J)+X(I)*PJ(I)*PJ(I)*W(I) 21 D(J)=D(J)+ER(I)*PJ(I)*W(I) [$\sum_n P_n(x_n) W(x_n)$ = $\sum [f_n - d_1 P_0 - d_2 P_1 - \dots] P_n W(x_n)$ — degree < n] D(J)=D(J)/S(J) DO 22 I=NP — $f_i = d_1 P_0 + d_2 P_1$ 22 ER(I)=ER(I)-D(J)*PJ(I) IF (J.EQ.NT) RETURN B(J)=B(J)/S(J) C(J)=S(J)/S(J-1) DO 27 I=1, NP P=PJ(I) PJ(I)=(X(I)-B(J))*PJ(I)-C(J)*PJM1(I) 27 PJM1(I)=P. GO TO 20 END </pre>
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Homework 11-2 4.11-3

Numerical Diff & Integration

an $f(x) \in$ smooth in some interval

$$f'(a) = \frac{f(a+h) - f(a)}{h} + \frac{1}{2} f''(a)h$$

$$f''(a) = \frac{f(a+h) - f(a-h)}{2h} + \frac{1}{6} f'''(\eta)h^2$$

$$f''(a) = \frac{f(a+h) - 2f(a) + f(a-h)}{h^2}$$

$$I(f) = \int_a^b f(x) dx$$

numerically integrate two types of formulas: local Ordinary & global Composite

$$\int_{-1}^1 \frac{x^2 dx}{1+x^2+x^4+\sin^5 x+11} = \sum_{i=1}^N \int_{x_i}^{x_{i+1}} () dx \quad -1 \leq x_1 < x_2 < \dots < x_{i+1} = +1$$

now apply ordinary numerical integration in $\int_{x_i}^{x_{i+1}} () dx$ Ordinary num. Integration in $[a, b]$ write $f(x) \approx P_k(x)$ k^{th} degree interp poly based on $x_0, x_1, x_2, \dots, x_k$
 $+ f[x_0, x_1, \dots, x_k] \psi_k(x) \quad (x-x_0)(x-x_1) \dots (x-x_k)$

$$I(f) = \int_a^b f(x) dx \approx I(P_k) + \int_a^b f[x_0, x_1, \dots, x_k, x] \psi_k(x) dx$$

Example $k=1$ $x_0=a$ $x_1=b$ $f(x) = f(a) + \frac{f(b)-f(a)}{b-a}(x-a) + f[a, b, x](x-a)(x-b)$

$$I(f) = f(a)(b-a) + \frac{f(b)-f(a)}{b-a} \frac{(b-a)^2}{2}$$

or, on the other hand, $f(x) \approx P_1(x) = \frac{f(b)-f(a)}{b-a}(x-a) + f(a)$

trapezoidal rule $(x-a)(x-b) < 0$

$$\text{error} = \frac{f''(\eta)}{2} \frac{(b-a)^3}{12}$$

for $n=2$ $x_0=a$ $x_1=\frac{a+b}{2}$ $x_2=b$

$$P_2(x) = f(a) + f[a, \frac{a+b}{2}](x-a) + f[a, \frac{a+b}{2}, b](x-a)(x-b)$$

$$I(f) = \int_a^b f(x) dx = \left[f(a) + \frac{1}{2} f(a+b) + f(b) \right] \frac{b-a}{6} - \frac{1}{90} (b-a)^5 f^{(4)}(\eta)$$

Gaussian Quadrature

$$f(x) = f(c) + f'(c)(x-c) + \dots + \frac{f^{(n)}(c)}{(n-1)!} \frac{(x-c)^{n-1}}{(n-1)!} + \int_c^x \frac{(x-s_{n-1})^{n-1}}{(n-1)!} f^{(n)}(s_{n-1}) ds_{n-1}$$

$$= \frac{f^{(n)}(c)}{n!} (x-c)^n + \frac{1}{n!} \int_c^x (x-s_n)^n f^{(n+1)}(s_n) ds_n$$

(A) but $\int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1$

when $f''(s_2) = f''(c) + \int_c^{s_2} f'''(s_3) ds_3$ is placed in it gives

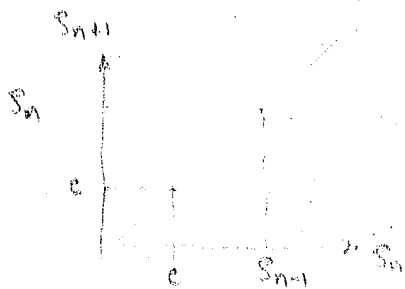
$$\int_c^x \int_c^{s_1} f''(c) ds_2 ds_1 + \int_c^x \int_c^{s_1} \int_c^{s_2} f'''(s_3) ds_3 ds_2 ds_1$$

we can by induction show that

$$f(x) = f(c) + \int_c^x f'(c) ds_1 + \int_c^x \int_c^{s_1} f''(c) ds_2 ds_1 + \dots + \int_c^x \int_c^{s_1} \dots \int_c^{s_{n-1}} f^{(n)}(c) ds_n ds_{n-1} \dots ds_1$$

$$+ \int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} \dots ds_1$$

$$f(c) + f'(c)(x-c) + \frac{f''(c)}{2}(x-c)^2 + \dots + \frac{f^{(n)}(c)}{n!} (x-c)^n + \int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} \dots ds_1$$



$$\int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n ds_{n-1} \dots ds_1$$

$$= \int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n ds_{n-1} \dots ds_1$$

$$f(x) = f(c) + \int_c^x f'(s_1) ds_1$$

$$= f(c) + \int_c^x f'(c) ds_1 + \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1$$

$$+ \int_c^x \int_c^{s_1} f''(c) ds_2 ds_1 + \int_c^x \int_c^{s_1} \int_c^{s_2} f'''(s_3) ds_3 ds_2 ds_1$$

Assume $m = n+1$
 $n = n$
 $1 = n-1$

by Taylor's expansion $f(x) = f(c) + f'(c)(x-c) + \frac{f''(c)}{2!}(x-c)^2 + \frac{f'''(c)}{3!}(x-c)^3 + \dots$
 $+ \frac{f^{(n)}(c)}{n!}(x-c)^n + \frac{1}{n!} \int_c^x f^{(n+1)}(s) (x-s)^n ds$ (1)

by the fundamental theorem of calculus

$$f(x) = f(c) + \int_c^x f'(s_1) ds_1 \quad (a)$$

$$f'(s_1) = f'(c) + \int_c^{s_1} f''(s_2) ds_2 \quad (b)$$

substitution of (b) into (a) gives

$$f(x) = f(c) + \int_c^x \left[f'(c) + \int_c^{s_1} f''(s_2) ds_2 \right] ds_1 \quad (c)$$

Since $f'(c)$ is a constant then (c) can be reduced to

$$\begin{aligned} f(x) &= f(c) + \int_c^x f'(c) [ds_1] + \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 \\ &= f(c) + f'(c)(x-c) + \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 \end{aligned}$$

Consider the domain



$$\int_c^x \int_c^{s_1} \int_c^{s_2} f'''(s_3) ds_3 ds_1 ds_2$$

$$\int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 = \int_c^x \int_{s_2}^x f''(s_2) ds_1 ds_2 = \int_c^x (x-s_2) f''(s_2) ds_2$$

by applying integration by parts

$$\int_c^x (x-s_2) f''(s_2) ds_2 = -\frac{(x-s_2)^2}{2} f''(s_2) \Big|_c^x + \int_c^x \frac{(x-s_2)^2}{2} f'''(s_2) ds_2$$

$$\begin{aligned} (x-s_2) ds_2 &= dv & v &= \frac{(x-s_2)^2}{2} \\ f'' &= u & du &= f''' ds_2 \end{aligned}$$

$$+ \frac{(x-c)^2}{2} f''(c) + \int_c^x \frac{(x-s_2)^2}{2} f'''(s_2) ds_2$$

$$f(x) = f(c) + f'(c)(x-c) + \frac{(x-c)^2}{2} f''(c) + \int_c^x \frac{(x-s_2)^2}{2} f'''(s_2) ds_2$$

$$f''(s_2) = f''(c) + \int_c^{s_2} f'''(s_3) ds_3 \quad \text{by induction}$$

$$f^{(n)}(s_n) = f^{(n)}(c) + \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1}$$

choose

$$x = \pm (.d_1 d_2 \dots d_n)_\beta \beta^e$$

$$f(x) = \pm (. \delta_1 \delta_2 \dots \delta_n)_\beta \beta^e$$

where $\delta_i = d_i$

$d_{n+1} \neq \delta_{n+1}$ since $\delta_{n+1} = 0$

$$x = - (.d_1 d_2 \dots d_n)_\beta \beta^e$$

$$f(x) =$$

1.32

$-x = \pm (.d_1 d_2 \dots d_n)_\beta \beta^e$ if only i digits are stored $i < n$

$fl(-x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e$ let $x_1 = |-x|$

$x_1 = \mp (.d_1 d_2 \dots d_n)_\beta \beta^e$ if only i digits are stored $i < n$

$fl(x) = \mp (.d_1 d_2 \dots d_i)_\beta \beta^e$

$-fl(x_1) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e = fl(-x)$

let $\beta^r x = \pm (.d_1 d_2 \dots d_n)_\beta \beta^{e+r}$ if only i digits are stored $i < n$

$fl(\beta^r x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^{e+r}$

$fl(x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e$ $i < n$

$\therefore \beta^r fl(x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e \cdot \beta^r = fl(\beta^r x)$

1.33

$fl(x)$ is given by chopping $fl(x) = x(1+\delta)$

assume $x^* = \pm (.d_1 d_2 \dots d_n)_\beta \beta^e$ assume i digits stored

$|x - x^*|$ where x^* is number by chopping x is "

$|x - x^*| = \pm (. \underbrace{00 \dots 0}_n d_{n+1} d_{n+2} \dots)_\beta \beta^e = \mp (. d_{n+1} d_{n+2} \dots)_\beta \beta^{e-n}$

$= \mp (. d_{n+1} d_{n+2} \dots)_\beta \pm (. d_1 d_2 \dots)_\beta \beta^e \beta^{-n}$
 $= \pm (. d_1 d_2 \dots)_\beta$

$= \pm (. d_{n+1} d_{n+2} \dots)_\beta \pm |x| \beta^{-n} \leq \pm (. \beta-1 \beta-1 \dots)_\beta |x| \beta^{-n}$
 $= \pm (. 1000 \dots)_\beta$

$\leq \beta |x| \beta^{-n} = |x| \beta^{1-n}$

$\frac{|x - x^*|}{x} \leq \beta^{1-n}$

1.4-1

$$x^2 + .4002 \cdot 10^9 x + 8 \cdot 10^{-4} = 0$$

$$x_1 = \frac{-.4002 \pm \sqrt{(.4002)^2 - 4(.8 \times 10^{-4})}}{2}$$

$$= \frac{-.4002 \pm \sqrt{.1598}}{2}$$

$$x_1 = \frac{-.4002 \pm .3999}{2} = \frac{-.0003}{2} = -.0002$$

$$\frac{-.00016}{.4002 \pm \sqrt{.1598}} = \frac{-.0002}{.8001} = -.0003$$

$$\begin{array}{r} .4002 \\ .4002 \\ \hline .16 \\ \hline .1602 \\ \hline .1601 \\ \hline .1602 \\ \hline .1601 \\ \hline .1602 \\ \hline .1601 \\ \hline .1602 \end{array}$$

$$\begin{array}{r} .4001 \\ .4001 \\ \hline .4001 \\ \hline .16004000 \\ \hline .16008001 \end{array}$$

$$\begin{array}{r} -.00032 \\ \hline .1598 \mid 400x \end{array}$$

$$\begin{array}{r} .39 \\ .39 \\ \hline .0351 \\ \hline .1170 \\ \hline .00035991 \\ \hline .000359910 \\ \hline .03599100 \\ \hline .11999000 \\ \hline .11999001 \end{array}$$

$$.4002 + .3999 = .8001$$

1.43

$$f(x) = \frac{\cos x(x) - \cos x \sin x}{\sin(x) \sin x} = x \cot x - \cos x$$

$$f(x) = \frac{x + \sqrt{x^2 - \alpha}}{x + \sqrt{x^2 - \alpha}} [x - \sqrt{x^2 - \alpha}] = \frac{x^2 - x^2 + \alpha}{x + \sqrt{x^2 - \alpha}} = \frac{\alpha}{x + \sqrt{x^2 - \alpha}}$$

1-76

$$\frac{h^{n+1}}{(n+1)!} f^{(n+1)}(c+\theta h) = \frac{h^{n+1}}{(n+1)!} f^{(n+1)}(c) + \epsilon$$

$$f^{(n+1)}(c+\theta h) = f^{(n+1)}(c) + \theta h f^{(n+2)}(c)$$

$$R_{n+1}(x) = \frac{1}{n!} \int_c^x (x-s)^n f^{(n+1)}(s) ds$$

$$= \frac{f^{(n+1)}(c) (x-c)^{n+1}}{(n+1)!} + \frac{1}{(n+1)!} \int_c^x (x-s)^{n+1} f^{(n+2)}(s) ds$$

$$= \frac{f^{(n+1)}(c) h^{n+1}}{(n+1)!} + \frac{(x-s)^{n+1}}{(n+1)!} \int_c^x f^{(n+2)}(s) ds \quad s \in [x, c]$$

$$\frac{(x-s)^{n+1}}{(n+1)!} f^{(n+1)}(x) - f^{(n+1)}(c) \leq \frac{h^{n+1}}{(n+1)!} [f^{(n+1)}(x) - f^{(n+1)}(c)]$$

$$\lim_{h \rightarrow 0} h^{n+1} \left[\frac{f^{(n+1)}(x) - f^{(n+1)}(c)}{(n+1)!} \right] \rightarrow 0 \Rightarrow o(h^{n+1})$$

$$x^2 - 2x - 2$$

$$\begin{array}{ll} x=1 & f=-3 \\ x=2 & f=-2 \\ x=3 & f=-1 \\ x=0 & f=-2 \\ x=-1 & f=1 \end{array}$$

$$x \in [-1, 0]$$

$$x \in [2, 3]$$

$$x = -0.5 \quad f = 4.25 + 1 - 2 = 3.25$$

$$x = -0.75 \quad f = \frac{9}{16} + \frac{3}{2} - 2 = \frac{1}{16}$$

$$x = -0.625 \quad f = \frac{25}{64} + \frac{80}{64} - \frac{128}{64} = \frac{-23}{64}$$

$$x = -0.6875$$

$$x = -0.71875$$

$$x = -0.734375$$

$$x = -0.7265625$$

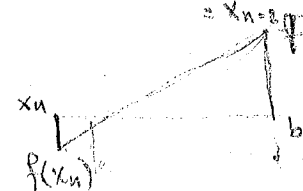
$$x_n = Fx_{n-1} - \frac{f(x_{n-1})}{f'(x_{n-1})}b$$

$$F = f'(x_n)$$

$$x_n - b = Fx_{n-1} - bF$$

$$\frac{x_n - b}{F} = x_{n-1} - b$$

$$\frac{x_n - b}{F^2} = \frac{x_{n-1} - b}{F} = x_{n-2} - b$$



$$x = \frac{1.0 - (-2) - 1}{1 + 2} = \frac{-2}{3} = -0.67$$

$$f = -2/3$$

$$x = \frac{1(-2/3) - (-2/3)(-1)}{1 + 2/3} = \frac{-8/9}{5/3} = -8/11 \approx -0.727$$

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

$$F = f'(x_n)$$

$$f(x_n) = \frac{F(\epsilon)}{F^{n+1}M}$$

$$f(x_{n+1} - x_n) = f(x_n)[x_{n+1} - b]$$

$$|x_1 - x_0| = \frac{M}{4}$$

$$|x_2 - x_1| = \frac{M}{8}$$

$$2^{-i} < \frac{4\epsilon}{M}$$

$$|x_{i+1} - x_i| = \frac{M}{2^{i+2}} < \epsilon$$

$$-i \log 2 < \log(\epsilon/M) + i \log 2$$

$$i > -2 - \frac{\log(\epsilon/M)}{\log 2}$$

$$\frac{.001}{1} = \log .001 = -3 \log 10$$

$$(b_n) b_n - f(b_n) b_n$$

$$s - x_{n+1} = s - \frac{f(b_n)a_n - f(a_n)b_n}{f(b_n) - f(a_n)} = \frac{f(b_n)[s - a_n] + f(a_n)[b_n - s]}{f(b_n) - f(a_n)}$$

$$x_{n+1} = \frac{f(b_n)[a_n - b_n] + b_n[f(a_n) - f(b_n)]}{f(b_n) - f(a_n)}$$

$$x_{n+1} - b_n = \frac{f(b_n)[a_n - b_n]}{f(b_n) - f(a_n)}$$

$$e_{n+1} = \frac{f(b_n) - f(a_n)}{f(b_n) - f(a_n)} \frac{a_n - b_n}{a_n - b_n}$$

constant

$$\rightarrow \frac{f'(s)}{f'(s)}$$

$$s \in (a_n, b_n)$$

$$i > \log \left[\frac{\epsilon}{M(F-1)} \right] / \log F$$

$$\text{where } F = f(b)$$

$$M = x_0 - b$$

2.2-2 1. $y = x^3 - 1$ funct at $x=0$ $f=1$ $x=0$ $f=1$ $f=1$ $x=1$

2. $g'(x) = 3x-1$ for $|x| \leq \frac{1}{3}$ $g'(x) \leq 1$

at $-\frac{1}{3}$ $-\frac{1}{3} - \frac{1}{3} - 1 < 1$ $\frac{27}{125} + \frac{3}{5} - 1$

$-\frac{1}{8} - \frac{1}{4} - 1$

$$x = \sqrt[3]{x+1} = g(x)$$

$$\frac{1}{3\sqrt[3]{(x+1)^2}} = g'(x) < 1 \quad 1 < x < 2$$

$$x = \tan x = g(x) \quad x=0$$

$$\sec^2 x = g'(x)$$

2 roots bet $0 \neq \pi/2$
 $0 \neq -\pi/2$

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

$$g(x) = \tan^{-1} x$$

$$g'(x) \leq 1$$

between $\pi/2$ & 0
 $-\pi/2$ & 0

$$e^{-x} = \cos x$$

$$g(x) = -x = -\ln(\cos x)$$

$$\frac{\sin x}{\cos x} = |\tan x| \leq 1$$

at $x = .707$ $g(x) = -\ln(.707) = -[\ln(.707) - \ln 1] = .303$

$$\log \ln a = \frac{\log_{10} a}{\log_{10} e} = \frac{\log_{10} e}{.434}$$

2.5-3

$$x = \pi/4$$

$$e = \frac{27}{5}$$

$$g(x) = x = \ln\left(\frac{1}{4}e^x\right)$$

$$\frac{27}{5} = \frac{1}{5}$$

$$g(x) = x = \ln(4 \cos x)$$

$$g'(x) = \frac{4 \sin x}{4 \cos x} = -\tan x$$

$$\cos(x) \leq 1$$

$$e^x + e^x e^{-x} = e^x e^{-x} + e^{-x}$$

$$\ln[e^x + e^{-x}]$$

$$f\left[\frac{(e^x + e^{-x})}{2}\right] = x$$

$$g(x) = x = \cosh^{-1}(2 \cos x)$$

bounded $[0, +2]$

$$0 < x < \pi/2$$

Prove:
$$\int_c^x \int_c^{s_1} \int_c^{s_2} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n \dots ds_1 = \frac{1}{n!} \int_c^x (x-s_2)^n f^{(n+1)}(s_2) ds_2$$

by fundamental theorem of calc:

$$f(x) = f(c) + \int_c^x f'(s_1) ds_1 \quad \text{and}$$

$$f'(s_1) = f'(c) + \int_c^{s_1} f''(s_2) ds_2$$

...

$$f^{(n)}(s_n) = f^{(n)}(c) + \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1}$$

this then leads to

$$f(x) = f(c) + \int_c^x f'(c) ds_1 + \int_c^x \int_c^{s_1} f''(c) ds_2 ds_1 + \dots + \int_c^x \int_c^{s_1} \dots \int_c^{s_{n-1}} f^{(n)}(c) ds_n ds_{n-1} \dots ds_1$$

$$+ \int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n \dots ds_1$$

since $f^{(k)}(c) = \text{constant}$ we can perform the first n integrals to produce

$$f(x) = f(c) + f'(c) \frac{(x-c)^2}{2} + \dots + f^{(n)}(c) \frac{(x-c)^n}{n!} +$$

$$\int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n \dots ds_1 \quad (a)$$

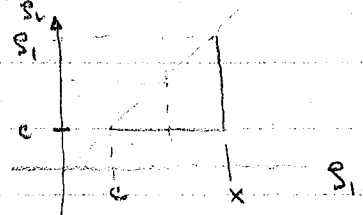
Again by fundamental theorem of calculus

$$f(x) = f(c) + \int_c^x f'(s_1) ds_1$$

Evaluate the integral by parts

$$\begin{aligned} f(x) &= f(c) + \int_c^x f'(s_1) ds_1 = f(c) + f'(c) \int_c^x ds_1 + \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 \\ &= f(c) + f'(c) (x-c) + \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 \end{aligned}$$

Consider the domain



$$\int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 = \int_c^x \int_{s_2}^x f''(s_2) ds_1 ds_2$$

$$\therefore \int_c^x \int_c^{s_1} f''(s_2) ds_2 ds_1 = \int_c^x (x - s_2) f''(s_2) ds_2$$

by limit interchange; thus by integration by parts

$$f(x) = f(c) + f'(c)(x-c) - \frac{(x-s_2)^2}{2} f''(s_2) \Big|_c^x + \int_c^x \frac{(x-s_2)^2}{2} f'''(s_2) ds_2$$

carrying out $n-2$ such partial integrations leads to

$$f(x) = f(c) + \sum_{i=1}^n \frac{f^{(i)}(c)(x-c)^i}{i!} + \frac{1}{n!} \int_c^x (x-s_2)^n f^{(n+1)}(s_2) ds_2 \quad (b)$$

if (a) & (b) are proper representations of the same function $f(x)$ then

$$\int_c^x \int_c^{s_1} \dots \int_c^{s_n} f^{(n+1)}(s_{n+1}) ds_{n+1} ds_n \dots ds_1 = \frac{1}{n!} \int_c^x f^{(n+1)}(s_2) (x-s_2)^n ds_2$$

1. Theorem: let $G \subset \mathbb{R}^N$ & let $F: G \rightarrow G$ for any $x_1, x_2 \in G$
and $\|F(x_1) - F(x_2)\| \leq K \|x_1 - x_2\|$ $K < 1$

Then for any $x_0 \in G$ the sequence of vectors x_n where $x_{n+1} = F(x_n)$
converges to a unique vector $\underline{x}$, i.e. $\|x_n - \underline{x}\| \rightarrow 0$ as $n \rightarrow \infty$
& $\underline{x} = F(\underline{x})$

Assume a sequence of vectors $x_1, x_2, \dots$ exist for $m \geq n$

$$\begin{aligned}\|x_m - x_n\| &= \|(x_m - x_{m-1}) + (x_{m-1} - x_{m-2}) + \dots + (x_{n+1} - x_n)\| \\ &\leq \|x_m - x_{m-1}\| + \|x_{m-1} - x_{m-2}\| + \dots + \|x_{n+1} - x_n\| \\ \|x_m - x_{m-1}\| &= \|F(x_{m-1}) - F(x_{m-2})\| \leq K \|x_{m-1} - x_{m-2}\| \\ &\leq K \|F(x_{m-2}) - F(x_{m-3})\| \leq K^2 \|x_{m-2} - x_{m-3}\| \\ &\dots \leq K^{m-1} \|x_1 - x_0\|\end{aligned}$$

in the same manner

$$\|x_m - x_n\| \leq K^n (1 + K + K^2 + \dots + K^{m-1-n}) \|x_1 - x_0\| \leq \frac{K^n}{1-K} \|x_1 - x_0\|$$

$$\text{if } K < 1 \text{ as } n \rightarrow \infty \quad K^n \rightarrow 0 \quad \therefore \|x_m - x_n\| \rightarrow 0 \quad \underline{x} = F(\underline{x})$$

$$\text{note } \|\underline{a} + \underline{b}\| = \max_i |a_i + b_i| \leq \max_i (|a_i| + |b_i|) \leq \max_i |a_i| + \max_i |b_i| = \|\underline{a}\| + \|\underline{b}\|$$

2. Theorem: If $\text{rank}(A) = r$ then $\exists$ exactly r l.i. row vectors of A
 $\Rightarrow \text{rank}(A) = \text{column rank}(A) = \text{row rank}(A)$

Suppose that the first r columns are linear indep. then

$$\sum_{i=1}^r c_i a_{ij} = 0 \iff c_1 = c_2 = \dots = c_r = 0, \quad j = 1, 2, 3, \dots, n$$

let us take n constants d_j s.t. we form the sum

$$\sum_{j=1}^n \sum_{i=1}^r d_j c_i a_{ij} = 0$$

Let us switch the sum signs $\therefore \sum_{i=1}^n c_i \sum_{j=1}^n d_j a_{ij} = 0$

$$\text{but } \sum_{j=1}^n d_j a_{ij} \Rightarrow \sum_{j=1}^r d_j a_{ij} + \sum_{j=r+1}^n d_j a_{ij}$$

since r a 's are l.i. then we can express $n-r$ vectors in terms

the r remaining l.i. vectors

because we can rewrite this sum as a $A\mathbf{d}$, where $A = (a)_{ij}$ & $\mathbf{d} = d_j$

the solution to $A\mathbf{d} = 0$ will have an ident matrix in the first r columns

since they are l.i. and the $n-r$ remaining a_{ij} can be expressed in terms
on the other r vectors, $\therefore$ if $\sum_{j=1}^n d_j a_{ij} = 0$ then $k = r$.

1.33 Let $-x = \pm (.d_1 d_2 \dots d_n)_\beta \beta^e$; if only i digits are stored $i < n$
 $fl(-x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e$ let $x_1 = -x$

$x_1 = \mp (.d_1 d_2 \dots d_n)_\beta \beta^e$; if only i digits are stored $i < n$

$$fl(x_1) = \mp (.d_1 d_2 \dots d_i)_\beta \beta^e$$

$$-fl(x_1) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e = fl(-x)$$

let $\beta^r x = \pm (.d_1 d_2 \dots d_n)_\beta \beta^{e+r}$ if only i digits are stored $i < n$

$$fl(\beta^r x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^{e+r}$$

$$fl(x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^e \quad i < n$$

$$\beta^r fl(x) = \pm (.d_1 d_2 \dots d_i)_\beta \beta^{e+r} = fl(\beta^r x)$$

1.33

let $fl(x)$ be given by chopping & $\delta = \delta(x) \ni fl(x) = x(1 + \delta)$

Assuming $x^* = \pm (.d_1 d_2 \dots d_n)_\beta \beta^e$ and i digits are stored

$$x - x^* = \mp (.000 \dots 0 d_{i+1} d_{i+2} \dots)_\beta \beta^e = \mp (.d_{i+1} d_{i+2} \dots)_\beta \beta^{e-i}$$

$$= \mp (.d_{i+1} d_{i+2} \dots)_\beta \beta^e \beta^{-i}$$

$$= \mp (.d_{i+1} d_{i+2} \dots)_\beta \beta^{-i}$$

$$= \mp (.d_{i+1} d_{i+2} \dots)_\beta x^* \beta^{-i}$$

$$= \pm (.d_{i+1} d_{i+2} \dots)_\beta \beta^{-i}$$

$$x - x^* = fl(x^*) - x^* = \delta x^* \quad \therefore \quad \delta = \mp (.d_{i+1} d_{i+2} \dots)_\beta \beta^{-i}$$

$$= \pm (.d_{i+1} d_{i+2} \dots)_\beta \beta^{-i}$$

let |denominator| be the lowest number permitted in the base > 0

let |numerator| be the largest " " in the bases $< \beta$

$$\delta \geq \mp (.(\beta-1)(\beta-1) \dots)_\beta \beta^{-i} \quad \text{note that this number is } \leq 0$$

$$= \pm (.100 \dots)_\beta \beta^{-i}$$

$$= \pm \beta^{1-i}$$

$$-\beta^{1-i} < \delta < 0$$

$$1.41 \quad x^2 + .4002x + .8 \cdot 10^{-4} = 0$$

$$x_1 = \frac{-.4002 \pm \sqrt{.4002^2 - 4(.8 \times 10^{-4})}}{2} = \frac{-.4002 \pm \sqrt{.1598}}{2} = \frac{-.0002}{2} = -.0001$$

$$x_1 = \frac{-.0002}{.4002 \pm \sqrt{.1598}} = \frac{-.0002}{.8001} = -.0003$$

$$1.43 \quad f(x) = \frac{x - \sin x}{\tan x} = \frac{x \cos x - \cos x \sin x}{\cos x} = \frac{x \cos x - .5 \sin 2x}{\cos x}$$

$\cos x$ acts as a magnifier on x whereas $\sin 2x \rightarrow 0$ much faster & denominator finite at $x \rightarrow 0$

$$f(x) = x - \sqrt{x^2 - a} = \frac{x^2 - x^2 + a}{x + \sqrt{x^2 - a}} = \frac{a}{x + \sqrt{x^2 - a}}$$

$$1.76 \quad R_{n+1}(x) = \frac{1}{n!} \int_c^x (x-s)^n f^{(n+1)}(s) ds$$

$$= \frac{f^{(n+1)}(c)}{(n+1)!} (x-c)^{n+1} + \frac{1}{(n+1)!} \int_c^x (x-s)^{n+1} f^{(n+2)}(s) ds$$

$$= \frac{f^{(n+1)}(c)}{(n+1)!} h^{n+1} + \frac{(x-s)^{n+1}}{(n+1)!} \int_c^x f^{(n+2)}(s) ds \quad \text{for } [x, c]$$

$$\frac{(x-s)^{n+1}}{(n+1)!} [f^{(n+1)}(x) - f^{(n+1)}(c)] \leq \frac{h^{n+1}}{(n+1)!} [f^{(n+1)}(x) - f^{(n+1)}(c)]$$

$$\lim_{h \rightarrow 0} h^{n+1} \left[\frac{f^{(n+1)}(x) - f^{(n+1)}(c)}{(n+1)!} \right] \rightarrow 0 \Rightarrow o(h^{n+1})$$

$$2.11 \quad \text{root} \in [2, 3]$$

x^* has 2 significant figures in about 10 iterations for

bisection, 5 iterations for regula falsi

$$\text{for bisection } |x_{i+1} - x_i| = \frac{M}{2^{i+2}} \leq \epsilon \rightarrow i \geq -2 - \frac{\log(\epsilon/M)}{\log 2}$$

$$\epsilon = 10^{-7} \quad M = 1$$

Assume $b_n \neq n$ remains const $\therefore$

$$\begin{aligned} x_{n+1} - x_n &= [F x_n - f(x_n) b] / (F - f(x_n)) - x_n \\ &= \frac{F x_n - f(x_n) b - F x_n + x_n f(x_n)}{F - f(x_n)} = \frac{f(x_n) [x_n - b]}{F - f(x_n)} \end{aligned}$$

$$\left| \frac{f(x_n) [x_n - b]}{F - f(x_n)} \right| < \epsilon \quad [F - f(x_n)] \epsilon > f(x_n) [x_n - b]$$

$$x_n - b = (F x_{n-1} - f(x_{n-1}) b) / (F - f(x_{n-1})) - b = F [x_{n-1} - b]$$

$$\Rightarrow \frac{x_n - b}{F^n} = x_0 - b = M \quad \text{where } x_0 \text{ is a}$$

$$[F - f(x_n)] \epsilon > f(x_n) F^n [x_0 - b]$$

$$F \epsilon - f(x_n) \epsilon > F^n f(x_n) M$$

$$\text{but } f(x_n) = \frac{F \epsilon}{x_{n+1} - b} = \frac{F \epsilon}{F^{n+1} M}$$

$$F \epsilon - \frac{\epsilon^2}{F^n M} > F^n \frac{F \epsilon}{F^{n+1} M} M > \epsilon$$

$$F - \epsilon / F^n M > 1$$

$$F^{n+1} M - \epsilon > F^n M$$

$$F^n (M - F M) < -\epsilon$$

$$F^n > \frac{\epsilon}{F M - M} \quad n > \log \left[\frac{\epsilon}{M(F-1)} \right] / \log F$$

$$\begin{aligned} 2.14 \quad |x_0 - a| &= \frac{(a-b)}{2} \quad |x_1 - x_0| = \frac{x_0 - a}{2} \quad \dots \quad |x_{i+1} - x_i| = \frac{|x_i - x_{i-1}|}{2} \\ |x_2 - x_1| &= \frac{x_1 - x_0}{2} \\ |x_{i+1} - x_i| &= \frac{|x_{i-1} - x_{i-2}|}{4} = \dots = \frac{|x_{i-k} - x_{i-(k+1)}|}{2^{k+1}} \quad \text{if } k = i-1 \end{aligned}$$

$$\frac{|x_1 - x_0|}{2^i} = |x_{i+1} - x_i| = \frac{x_0 - a}{2^{i+1}} = \frac{(a-b)}{2^{i+2}} = \frac{M}{2^{i+2}}$$

$$2.23 \quad e^x = 1 + x + \frac{x^2}{2} + \frac{1}{2!} \int_0^x (x-s)^2 e^s ds = 1 + x + \frac{x^2}{2} + \frac{e^s}{3!} (x)^3 \quad 0 \leq s \leq x$$

$$f(x) = e^x - 1 - x - \frac{x^2}{2} = \frac{e^s}{3!} x^3 \quad f(x) \rightarrow 0 \quad \text{where } x \rightarrow 0$$

2.3-2

for (a) root $\in [1, 2]$ take $x = \sqrt[3]{x+1}$

since $g'(x) = \frac{1}{3}(x+1)^{-2/3} < 1$ in I

for (b) root $\in [\frac{\pi}{4}, \frac{\pi}{2}]$ take $x = \tan^{-1}(x)$

since $g'(x) = \frac{1}{1+x^2} < 1$ in I

for c root $\in [\frac{\pi}{4}, \frac{3\pi}{4}]$ take $x = \arccos(e^{-x})$

since $g'(x) = \frac{1}{\sqrt{1+e^{-2x}}} < 1$ in I

2.5-3 let $x=0$ $f(x) = 4\cos x - e^x$ $f'(x) = -4\sin x - e^x$

root lies between 0 & 1 $f'(x)$ varies between -1 & -6 in the interval

$f''(x) = -4\cos x - e^x$; f, f', f'' are continuous fns in the interval

$\exists 0 < \xi < 1$ $\therefore f(\xi) = 0$ and $-6 < f'(\xi) < -1$

let $x=1$ $f(x) = 2\cos x - \cosh x$ $f'(x) = -2\sin x - \sinh x$

root lies between $\frac{1}{2}$ & 1 $f'(x)$ varies between -1.4 & -2.7

$f''(x) = -2\cos x - \cosh x$; f, f', f'' are continuous fns in this

interval. $\exists \frac{1}{2} < \xi < 1$ $\therefore f(\xi) = 0$ & $-2.7 < f'(\xi) < -1.4$

let $x = \frac{\pi}{4}$ $f(x) = e^{-x^2} - \cos x$ $f'(x) = -2xe^{-x^2} + \sin x$

root lies between $\frac{\pi}{4}$ & $\frac{3\pi}{4}$ $f'(x)$ varies between -9 & -0.5

$f''(x) = +4x^2e^{-x^2} - 2e^{-x^2} + \cos x$; f, f', f'' are cont. fns in this

interval, $\exists \frac{\pi}{4} < \xi < \frac{3\pi}{4}$ $\therefore f(\xi) = 0$ & $-9 < f'(\xi) < -0.5$

```

000003      PROGRAM MULLER(INPUT,OUTPUT,TAPE5=INPUT,TAPE6=OUTPUT)
000003      EXTERNAL FN
000003      COMPLEX FM,XM,FN
000003      COMMON M
000003      DIMENSION A(5),FM(3),XM(3)
000003      READ(5,1) M,NDEG
000013      1  FORMAT(2I5)
000013      COMMENT N IS THE ACTUAL DEGREE OF POLYNOMIAL, I.E. X**3 IS A THIRD DEGREE
000013      COMMENT THEREFORE N=3
000013      COMMENT M IS THE NUMBER OF ITERATIONS ALLOWED BEFORE EXITING ON NONCONVER
000013      N=NDEG+1
000013      COMMENT N IS THE DEGREE OF THE POLYNOMIAL + 1 TO ACCOMODATE THE ARRAY OF
000013      COMMENT COMPUTER SINCE A0 IS A(1)....
000015      READ(5,2) (A(I),I=1,N)
000030      2  FORMAT(8F10,0)
000030      DO 10 I=1,3
000032      FM(I)=CMPLX(0.000,0.000)
000040      10  XM(I)=CMPLX(0.000,0.000)
000051      DO 400 I=1,21
000052      AK=(1-I)*0.1
000055      C      GENERATION OF INITIAL DATA
000060      C      XM(I) I=1,3 ARE X(I-2),X(I-1),X(I)
000063      C      FM(I) I=1,3 ARE F(I-2),F(I-1),F(I)
000065      XM(1)=CMPLX(0.5,-1.7)
000071      XM(2)=CMPLX(0.5,-1.8)
000075      XM(3)=CMPLX(0.5,-2.1)
000101      FM(1)=FN(XM(1),AK,A,N)
000107      FM(2)=FN(XM(2),AK,A,N)
000113      FM(3)=FN(XM(3),AK,A,N)
000121      IF(I.EQ.1) WRITE(6,15)
000123      15  FORMAT(1H1,14X,#REAL#,12X,#IMAG#,18X,#REAL#,12X,#IMAG#,/)
000124      CALL MULL(XM,FM,FN,A,AK,N)
000125      IF(AK*NE,1.00000) WRITE(6,20)
000126      20  FORMAT(1X,/)
000127      400 CONTINUE
000128      CALL EXIT
000129      END

```


RUN VERSION 2,3 - CDC LEVEL 26 - NYU LEVEL 11

MULLER

PROGRAM LENGTH INCLUDING I/O BUFFERS
004303

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1	-	000137	2	-	000141	15	-	000155	20	-	000165
400	-	000121									

BLOCK NAMES AND LENGTHS
- 000001

VARIABLE ASSIGNMENTS

A	-	000226	AK	-	000236	FM	-	000212	I	-	000235
M	-	000000C01	N	-	000234	NDEG	-	000233	XM	-	000220

START OF CONSTANTS
000126

START OF TEMPORARIES
000167

START OF INDIRECTS
000207

UNUSED COMPILER SPACE
013000



```

SUBROUTINE MULL(X,FX,G,A,AK,NDEG)
COMPLEX X,FX,G,EPSI,HI,ALI,FI,SQ,ANU,ADE,XIP1,FXIP1
COMMON M
DIMENSION X(3),FX(3),A(NDEG)
EPSI=CMPLX(1.0E-04,1.0E-04)
J=1
IFLAG=0
5 I=1
HI=X(3)-X(2)
ALI=HI/(X(2)-X(1))
10 FI=(1.+2.*ALI)*(FX(3)-FX(2))-ALI*ALI*(FX(2)-FX(1))
SQ=FI*FI-4.*FX(3)*(ALI+ALI*ALI)*(FX(3)-FX(2)-ALI*(FX(2)-FX(1)))
ANU=-2.*FX(3)*(1.+ALI)
ADE=FI+CSQRT(SQ)
IF(CABS(FI-CSQRT(SQ)).GT,CABS(ADE)) ADE=FI-CSQRT(SQ)
ALI=ANU/ADE
XIP1=X(3)+HI*ALI
HI=HI+ALI
FXIP1=G(XIP1,AK,A,NDEG)
FX(1)=FX(2)
FX(2)=FX(3)
FX(3)=FXIP1
X(1)=X(2)
X(2)=X(3)
X(3)=XIP1
IF(CABS(X(3)-X(2)).LE,CABS(EPSI)) GO TO 20
I=I+1
IF(I.GT. M) GO TO 22
GO TO 10
20 IF(CABS(FX(3)).LE,CABS(EPSI)) GO TO 30
22 WRITE(6,25) X(3),FX(3),I,AK
25 FORMAT(/,5X,'ROOT CONVERGENCE #,E14,7,2X,E14,7,'NOT FNL CONVERGE#
1,'NCE #,E14,7,2X,E14,7,/,5X,'ITER COUNT= #,I5,' FOR A K= #,F10,5)
CALL EXIT
30 IF(ABS(AIMAG(X(3))).LT,AIMAG(EPSI)) GO TO 50
WRITE(6,35) X(3),FX(3),I,AK
35 FORMAT(5X,'X=#,2(2X,E14,7),# FX=#,2(2X,E14,7),# ITER CNT=#,I3,
1# K=#,F10,5)
AX=REAL(X(3))
BX=-AIMAG(X(3))
X(3)=CMPLX(AX,BX)
FX(3)=G(X(3),AK,A,NDEG)
WRITE(6,40) X(3),FX(3)
40 FORMAT(5X,'X=#,2(2X,E14,7),# FX=#,2(2X,E14,7))
IFLAG=IFLAG+1
J=J+2*(IFLAG)
IF(J.EQ,NDEG) GO TO 100
C HERE WOULD BE PLACED SOME METHOD OF DETERMINING NEXT ROOT APPROX.
000533 X(1)=CMPLX(-0.3,0.000)
000540 X(2)=CMPLX(-0.4,0.000)
000546 X(3)=CMPLX(-0.45,0.000)
000553 FX(1)=G(X(1),AK,A,NDEG)
000572 FX(2)=G(X(2),AK,A,NDEG)
000611 FX(3)=G(X(3),AK,A,NDEG)
000630 GO TO 5
000630 50 WRITE(6,35) X(3),FX(3),I,AK
000654 J=J+1

```



```

000656      IF(J.EQ,NDEG) GO TO 100
C          HERE WOULD BE PLACED SOME METHOD OF DETERMINING NEXT ROOT APPROX,
C          I.E. DEFLATION
000663      100 IF(AK.NE,1,00000) RETURN
000666          N=J-1+2*IFLAG
000671          WRITE(6,110) IFLAG,N
000701      110 FORMAT(//,10X,*,THERE ARE *,13,*,PAIRS OF COMPLEX ROOTS AND *,13,
          1*,REAL ROOTS*,//)
000701          RETURN
000702          END

```

C

C

MULL

SUBPROGRAM LENGTH
001110

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS											
5	-	000015	10	-	000040	20	-	000343	22	-	000363
25	-	000714	30	-	000410	35	-	000732	40	-	000743
50	-	000630	100	-	000663	110	-	000757			

BLOCK NAMES AND LENGTHS
- 000001

VARIABLE ASSIGNMENTS											
ADE	-	001074	ALI	-	001064	ANU	-	001072	AX	-	001105
BX	-	001106	EPSI	-	001060	FI	-	001066	FXIP1	-	001100
HI	-	001062	J	-	001104	IFLAG	-	001103	J	-	001102
M	-	000000C01	N	-	001107	SG	-	001070	XIP1	-	001076

START OF CONSTANTS
000704

START OF TEMPORARIES
000770

START OF INDIRECTS
001034

UNUSED COMPILER SPACE
010600


```

000007 COMPLEX FUNCTION FN(X,AK,A,N)
000007 COMPLEX X,B,CHTR,CAK
000007 DIMENSION A(N)
000007 B=CMPLX(A(N),0,000)
000012 NM1=N*1
000014 DO 10 K=1,NM1
000015 KK=NM1-K*1
000017 CHTR=CMPLX(A(KK),0,000)
000023 10 B=CHTR*X*B
000036 CAK=CMPLX(AK,0,000)
000040 FN=B+CAK
000045 RETURN
000047 END

```

C

C

FN

SUBPROGRAM LENGTH
000100

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS
B - 000067 OAK - 000073 CHTR - 000071 FN - 000065
K - 000076 KK - 000077 NM1 - 000075

START OF CONSTANTS
000051

START OF TEMPORARIES
000055

START OF INDIRECTS
000064

UNUSED COMPILER SPACE
013300

我欲與君共事，君必欲先見我。我欲與君共事，君必欲先見我。


```

PROGRAM STEFF(INPUT,OUTPUT,TAPE5=INPUT,TAPE6=OUTPUT)
C PROGRAM TO FIND X SUCH THAT X=G(X)
000003 WRITE(6,1)
000007 1 FORMAT(1H1,4X,*,STEFFENSEN ITERATION TECHNIQUE*,//)
000007 I=1
000010 5 READ(5,3) X0
000016 3 FORMAT(F15,10)
000016 WRITE(6,4) X0
000024 4 FORMAT(//,5X,18HSTARTING POINT IS ,E14.7)
000024 ITER=1
000025 Y0=G(X0)
000030 10 Y1=G(Y0)
000032 Y2=G(Y1)
C OBTAIN DELTA1
000034 DL=Y2-Y1
C CHECK TO SEE IF DELTA1 IS LESS THAN EPSILON...IF YES WE HAVE RESULTS
000036 IF(ABS(DL),LE,1.0E-07) GO TO 100
C OBTAIN THE INVERSE OF THE DERIVATIVE
000042 RINV=(Y1-Y0)/DL
C IF THE DERIVATIVE IS GREATER THAN 1 USE AITKEN'S DELTA SQUARED
C PROCESS...IF NOT, START STEFFENSEN'S ITERATION
000044 IF(ABS(RINV),LE,1.0) GO TO 50
000050 NITER=ITER
C OBTAIN THE NEW GUESS AT THE ROOT OF THE FUNCTION
000051 Y0=Y2-DL/(RINV-1.)
C CHECK TO SEE IF THE ITERATION PROCEDURE CONVERGES...IF NOT STOP
000055 40 IF(ITER,GE,50) GO TO 75
000060 ITER=ITER+1
000061 GO TO 10
000062 50 Y0=Y2
000064 GO TO 40
C PRINT OUT THE VALUES WHEN ITERATION COUNT WAS EXCEEDED...THEN STOP
000064 75 GP=1./RINV
000066 WRITE(6,80) ITER,Y2,Y1,Y0,GP
000104 80 FORMAT(5X,*,NO CONVERGENCE AFTER*,I3,*, ITERATIONS*,/,5X,*,Y2=*,E14.7
1,2X,*,Y1=*,E14.7,2X,*,Y0=*,E14.7,2X,*,GPRIME=*,E14.7)
000104 CALL EXIT
C PRINT OUT THE RESULTS IF THE ITERATION PROCEDURE CONVERGED
000105 100 GP=1./RINV
C PRINT OUT THE NUMBER OF TIMES AITKENS WAS USED
000107 WRITE(6,2) NITER
000115 2 FORMAT(5X,*,GPRIME WAS ,GE. 1 FOR *,I3,*, ITERATIONS PRIOR TO CONFO*,
1,*,RMING TO CONDITIONS FOR CONVERGENCE*)
000115 WRITE(6,101) Y2,Y1,Y0,ITER,GP
000133 101 FORMAT(5X,*,Y2=*,E14.7,2X,*,Y1=*,E14.7,2X,*,Y0=*,E14.7,/,5X,*,CONVERG*,
1,*,ED AFTER*,I3,*, ITERATIONS WITH GPRIME EQUAL TO*,E14.7)
000133 I=I+1
000135 IF(I-5) 5,5,102
000137 102 CALL EXIT
000140 END

```


RUN VERSION 2,3 - CDC LEVEL 26 - NYU LEVEL 11

STEFF

PROGRAM LENGTH INCLUDING I/O BUFFERS

004335

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

1	-	000153	2	-	000210	3	-	000162	4	-	000164
3	-	000010	10	-	000030	40	-	000055	50	-	000062
75	-	000064	80	-	000173	100	-	000105	101	-	000224
102	-	000137									

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

DL	-	000265	GP	-	000270	I	-	000257	ITER	-	000261
NITER	-	000267	RINV	-	000266	XC	-	000260	YO	-	000262
Y1	-	000263	Y2	-	000264						

START OF CONSTANTS

000142

START OF TEMPORARIES

000241

START OF INDIRECTS

000257

UNUSED COMPILER SPACE

002700



```
000003 FUNCTION G(TSI)
000010 G=0,5+0,2*SIN(TSI)
000010 RETURN
000010 END
```


RUN VERSION 2.3 - CDC LEVEL 26 - NYU LEVEL 11

3

SUBPROGRAM LENGTH

000022

FUNCTION ASSIGNMENTS

STATEMENT ASSIGNMENTS

BLOCK NAMES AND LENGTHS

VARIABLE ASSIGNMENTS

3 - 000021

START OF CONSTANTS

000012

START OF TEMPORARIES

000015

START OF INDIRECTS

000021

UNUSED COMPILER SPACE

003400

C

C

CORE MAP 16,34,13, NORMAL CONTROL
 ---TIME---LOAD MCDE ---L1---L2---TYPE-----USEP-----CALL-----
 FWA LOADER 033771 FWA TABLES 032162
 -PROGRAM---ADDRESS- --LABELED---COMMON--
 STEFF 000100
 G 004435
 SINCOS 004457
 SYSTEM 004556 SCOPE2 004556
 GETBA 005662
 INPUTC 005701
 KODER 006031
 KRAKER 007354
 OUTPTC 010474
 SIOS 010566
 ---UNSATISFIED EXTERNALS----- REFERENCES



Jan 8, 1972

Exam: Jan 22 11:30 - 1:30 T2-332

Review of Num. Methods

1. Floating pt representation of real numbers

Rounding & Chopping methods. Bounds on Relative Error

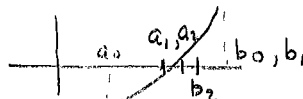
(Advantages of Rounding & Chopping)

Propagation of error & loss of significant figure in operation with floating pt. #'s.

Definition & familiarity with order symbols. $O()$ & $o()$

2. Soln of Non-linear Eqs.

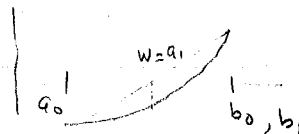
(1) Bisection



Given $[a_0, b_0]$ $f(a_0)f(b_0) < 0$

(2) Regular Falsi

$[a_0, b_0]$ $f(a_0)f(b_0) < 0$



(3) Modified Regular Falsi

$[a_0, b_0]$ $f(a_0)f(b_0) < 0$

Algorithm for mod regula falsi

Set $f = f(a_0)$, $g = f(b_0)$

$w_0 = a_0$

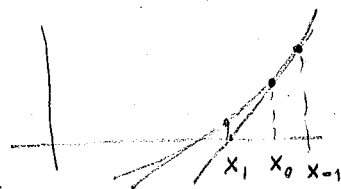
Do $n=0, 1, \dots$

Compute $w_{n+1} = \frac{Fb_n - Ga_n}{F - G}$

[if $f(a_n)f(w_{n+1}) \leq 0$, $a_{n+1} = a_n$ $b_{n+1} = w_{n+1}$ $G = f(w_{n+1})$
" " > 0 , $a_{n+1} = w_{n+1}$ $b_{n+1} = b_n$ $F = f(w_{n+1})$
If also $f(w_n)f(w_{n+1}) > 0$ set $F = F/2$
If also $f(w_n)f(w_{n+1}) > 0$ set $G = G/2$

(4) Secant method

$$[x_{-1}, x_0] \quad f(x)$$



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

do not know if root will be encountered
since don't know root lies in interval

(5) Newton's method

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

limit of (4) must know deriv.

Fixed Pt iteration

$$f(x)=0 \iff x=g(x)$$

$$g(x) = x - \frac{f}{f'} \text{ for Newton's method.}$$

General Result:

Given x_0 , $x_{n+1} = g(x_n)$

$x_{n+1} \rightarrow \xi$ where $\xi = g(\xi)$; if (i) $g: [a, b] \rightarrow [a, b]$

ξ is unique

(ii) $|g'(x)| \leq K < 1$ $x \in [a, b]$

Quest: describe fixed pt iteration using Newton's function for the evaluation of $\sqrt{2}$. Specify $[a, b]$ & check (ii)

Error in fixed pt iteration $x_{n+1} = g(x_n)$

$$e_n = \xi - x_n \quad e_{n+1} \sim g'(\xi)e_n - \frac{1}{2}g''(\xi)e_n^2 + \dots \quad n \rightarrow \infty$$

if $g'(\xi) \neq 0$ linear conv.

$g'(\xi) = 0$ quad.

On adjusting $g(x)$ so that $g'(\xi) = 0$ we obtain $g(x) = x - \frac{f(x)}{f'(x)}$

$\therefore$ Newton's method converges quad

Q: What happens if $f'(\xi) = 0$ in Newton's method
 Newton linear put $\frac{2f}{f'}$ Newton $\rightarrow$ guard.

Acceleration of convergence in fixed pt iteration $x_{n+1} = g(x_n)$ $\xi = g(\xi)$

Construct $x_0, x_1, \dots, x_n \rightarrow \xi$

$$e_n = \xi - x_n \quad x_n = \xi + O(e_n)$$

construct $\hat{x}_0, \hat{x}_1, \dots, \hat{x}_n \Rightarrow \hat{x}_n = \xi + o(e_n)$

Aitken's Δ^2 method $\hat{x}_n = x_{n+1} - \frac{(\Delta x_n)^2}{\Delta^2 x_{n-1}}$ $\Delta x_n = x_{n+1} - x_n$

Be able to argue through this formulae:

$$\xi - x_{n+1} \sim g'(\xi)(\xi - x_n)$$

On using Aitken in conj with $x_{n+1} = g(x_n) \rightarrow$ Steffensen's scheme

Set $y_0 = x_0$ Given

DO $n=0, 1, \dots$

$$\left. \begin{array}{l} x_1 = g(x_0) \\ x_2 = g(x_1) \end{array} \right\} \begin{array}{l} \Delta x_1 = x_2 - x_1 \\ \Delta^2 x_0 \end{array}$$

$$\text{Calculate } y_{n+1} = x_2 - \frac{(\Delta x_1)^2}{\Delta^2 x_0}$$

Polynomials: (i) Nested Multiplication (Synthetic Div)
 (ii) Newton's Method for real zeroes.

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

(iii) Muller's Method: Analogous to secant method except pass parabola through 3-pt.

1st Why is Muller's Method good? (Nashson & Keller)

Compare Newton & Muller's Method

3. Linear Eq's:

1. Elementary linear Algebra
2. The Alternative Theorem. *

When is $AX=b$ Solvable? if $AX=0 \Leftrightarrow X=0$ then $AX=b$ has! solut.
 otherwise $\exists$ vector $\Rightarrow$ if $AX=0$ $x = \sum_{j=1}^n \alpha_j x_j$

3. Back Substitution for Triangular system.
4. Ordinary Gauss elimination.
5. Gauss Elimination with max. partial pivot.

(describe: Work out an example)

Purpose of pivoting No pivoting $\Rightarrow$ inaccurate results.

2nd (*) 6. Describes $PLU=A$ factorization

Assuming you have G.Elim available.

7). Matrix Norms & Condition Numbers.

$$e = x - \hat{x} \quad \text{--- Compute}$$

$$r = Ae \quad \frac{1}{\|A\| \|A^{-1}\| \|b\| \|x\|} \|r\| \leq \|e\| \leq \|A\| \|A^{-1}\| \frac{\|r\|}{\|b\|}$$

$$\text{Cond} = \|A\| \|A^{-1}\|$$

8) Remarks on Condition Numbers.

3 (*) 9) Describe iterative methods for lin System.

$$A = \hat{L} + \hat{U} + \hat{D}$$

Jacobi

G.S.

Iterative Correct

Solve $Ax - b = 0$

$$x = x + C^{-1}(b - Ax) \Leftrightarrow Ax - b = 0 \quad C = \text{nothing}$$

$$x = (I - C^{-1}A)x + C^{-1}b$$

Jacobi $C = \hat{D}$ if $\hat{D} = (0)$ permute rows $\Rightarrow \hat{D} \neq (0)$

$$x_{n+1} = Bx_n + c \quad x_n \xrightarrow{?} \xi$$

G.S. $C = \hat{D} + \hat{L}$

(10) When does $x_{n+1} = Bx_n + c \Rightarrow$ convergence

Ans $\|x_n - \xi\| \rightarrow 0$ as $n \rightarrow \infty$ $\xi = B\xi + c \Leftrightarrow$ ^{mod} λ E.V. of B are less

i.e. $Bz = \lambda z \quad z \neq 0 \quad |\lambda| < 1$

Sufficient and for Cons. Jacobi: if A diag dominant $a_{kk} > \sum_{j \neq k} a_{kj}$
 $\|I - C^{-1}A\|$ has small norm

G.S.: same as Jacobi

11) Power method for eigenvalues $\lambda_1 > \lambda_2 > \dots > \lambda_n$

$B \quad Bz, B^2z, \dots$

$$\lambda_1 \sim \frac{j^{\text{th entry}} Bz^{m+1}}{j^{\text{th entry}} Bz^m}$$

Why is this OK.

Iterative Correction: up to us.

4. Interpolation & least Squares.

1) what is interpolation problem: given $x_0, x_1, \dots, x_n$ & $f_0, f_1, \dots, f_n$

Find a polyn $p_n(x)$ degree $\leq n \Rightarrow p_n(x_k) = f(x_k) \quad k=0, \dots, n$

2) Lagrange form of $P_n(x) = \sum_{k=0}^n f(x_k) \prod_{\substack{j=0 \\ j \neq k}}^n \frac{(x-x_j)}{(x_k-x_j)}$

3) Newton's Form

$$P_n(x) = f[x_0] + f[x_0, x_1](x-x_0) + \dots + f[x_0, \dots, x_n](x-x_0) \dots (x-x_{n-1})$$

$$f[x_0, x_1, \dots, x_n] = \frac{f[x_1, \dots, x_n] - f[x_0, \dots, x_{n-1}]}{x_n - x_0}$$

4) Error of $P_n(x)$

5) Piecewise poly interpolation cubic interpolation

Q: describe piecewise cubic poly interpolation

- (i) Bessel
- (ii) Hermite
- (iii) Spline

Comments on all three

$$\begin{array}{ccccccc} | & | & & | & & | & | \\ x_0 & x_1 & & x_{n-1} & & x_n & \\ & & \text{Knots} & & & & \end{array}$$

$$P_k(x) = c_0^k + c_1^k x + c_2^k x^2 + c_3^k x^3$$

$$x_n \rightarrow x_{n+1}$$

$$\begin{array}{lcl} \text{if } P_k(x_k) = f(x_k) & ; & P'_k(x_k) = P'_{k-1}(x_k) \\ P_k(x_{k+1}) = f(x_{k+1}) & ; & P'_k(x_{k+1}) = P'_{k+1}(x_{k+1}) \\ P'_k(x_k) = f'(x_k) & \downarrow & \end{array}$$

Hermite: f & deriv are needed at $x_0, x_1, \dots, x_n$ use limit of interp pts

Bessel: f' not given, same as above but $f'_k \sim \frac{\Delta f}{\Delta x}$

Spline: Adjust $c_j^k \Rightarrow P_k(x_k) = f(x_k)$

$$P_{k+1}(x_{k+1}) = f(x_{k+1})$$

other constants are determined by requiring that P.Wisc poly has 2 continuous derivs.

5. Least Squares

(i) State problem

(ii) deriv normal eqs.

(iii) Inner product & orthog pol def. Examples.

(iv) Major How does one use orthog poly for the numerical implementation of least square data fitting method: easy to program
recursiveness

(4/4)

C. Leung
763,1050
Prof. Papanicolaou

3.91 if λ is an eigenvalue of A $\exists$ an $\underline{x} \neq 0$ s.t. $A\underline{x} = \lambda\underline{x}$

if a is a scalar then $aA\underline{x} = a\lambda\underline{x}$

if b is a scalar then $bI\underline{x} = b\underline{x}$

then $aA\underline{x} + bI\underline{x} = a\lambda\underline{x} + b\underline{x}$

using distributive law of addition over multiplication

$$(aA + bI)\underline{x} = aA\underline{x} + bI\underline{x}$$

$$(a\lambda + b)\underline{x} = a\lambda\underline{x} + b\underline{x}$$

$$\therefore (aA + bI)\underline{x} = (a\lambda + b)\underline{x}$$

denote $aA + bI$ by B and $a\lambda + b$ by μ

then $B\underline{x} = \mu\underline{x}$ where B is a matrix (square) & μ is a scalar; then μ is an eigenvalue of B since $\exists$ an $\underline{x} \neq 0$

$$\Rightarrow (B - \mu I)\underline{x} = 0$$

$$3.92 \quad p(A) = c_0 A^0 + c_1 A^1 + c_2 A^2 + \dots + c_k A^k$$

where $A^0 = I$ and $A^j = A(A^{j-1})$

post multiply $p(A)$ by a non zero vector $\underline{x}$ which satisfies $A\underline{x} = \lambda\underline{x}$;
then $p(A)\underline{x} = \sum_{i=0}^k c_i A^i \underline{x} = c_0 I\underline{x} + \sum_{i=1}^k c_i A^{i-1} A\underline{x} = c_0 I\underline{x} + \sum_{i=1}^k c_i A^{i-1} \lambda\underline{x}$

$$= c_0 I\underline{x} + c_1 \lambda I\underline{x} + \lambda \sum_{i=2}^k c_i A^{i-2} A\underline{x} = c_0 I\underline{x} + c_1 \lambda I\underline{x} + \sum_{i=2}^k c_i \lambda^2 A^{i-2} \underline{x}$$

$$= \dots = c_0 I\underline{x} + c_1 \lambda I\underline{x} + c_2 \lambda^2 I\underline{x} + \dots + c_k \lambda^k I\underline{x} = p(\lambda) I\underline{x}$$

where $p(\lambda) = \sum_{i=0}^k c_i \lambda^i$; then $p(A)\underline{x} = p(\lambda) I\underline{x} = p(\lambda)\underline{x}$

let

B be a matrix $= p(A)$ and $\mu = p(\lambda)$ a scalar then

let A be tridiag $\Rightarrow a_{ii} = 0$ $a_{i,i+1} = a_{i+1,i}$ therefore

$$\begin{pmatrix} 0 & r_1 & & \\ r_1 & & & \\ & & & \\ & & & r_{n-1} \\ & & r_{n-1} & 0 \end{pmatrix}$$

then $\exists$ a non zero $x \Rightarrow Ax = \lambda x$

let $x_i = \sin(i\pi/(n+1))$ for an $n \times n$ matrix, argument of sin $0 < \frac{i\pi}{n+1} < \pi$

$\therefore$

$$(A - \lambda I)x = 0 \text{ iff } \det(A - \lambda I) = 0$$

$$A - \lambda I = \begin{pmatrix} -\lambda & r_1 & & \\ & r_1 & & \\ & & & r_{n-1} \\ & & r_{n-1} & -\lambda \end{pmatrix}$$

write $A - \lambda I$ as LU where L is a lower triang matrix & U is upper

$$L = \begin{pmatrix} \alpha_1 & & & \\ \beta_2 & \alpha_2 & & \\ & & \ddots & \\ 0 & & & \beta_n & \alpha_n \end{pmatrix} \quad U = \begin{pmatrix} 1 & \gamma_1 & & \\ & \ddots & & \\ & & \gamma_{n-1} & \\ & & & 1 \end{pmatrix}$$

where $\alpha_i, \beta_i, \gamma_i$ are defined by

$$\alpha_1 = -\lambda \quad \beta_i = r_{i-1} \quad \gamma_i = r_i / -\lambda$$

$$\text{and } \alpha_i = -\lambda - \beta_i \gamma_{i-1} \Rightarrow \alpha_i = -\lambda^2 + r_{i-1}^2$$

$$\det U = 1 \quad \det L = \alpha_1 \alpha_2 \dots \alpha_n = \frac{-\lambda(-\lambda^2 + r_1^2) \dots (-\lambda^2 + r_{n-1}^2)}{\lambda^{n-1}}$$

if λ is non zero

$$\det L = 0 = - \frac{(r_1^2 - \lambda^2) \dots (r_{n-1}^2 - \lambda^2)}{\lambda^{n-1}} = 0 \Rightarrow \lambda = r_i \quad (i=1, \dots, n-1)$$

but $r_1 = r_2 = \dots = r_{n-1} = 1$ then $\lambda = \pm 1$ of LU

take a look at i^{th} equation of j^{th} eigenvector

$$\sin(\underline{i-1})j\pi + \sin(\underline{i+1})j\pi = \lambda^j \sin \underline{i}j\pi$$

$$\text{but } \sin \frac{(i+1)j\pi}{n+1} = \lambda \frac{\sin ij\pi}{n+1} \cos \frac{j\pi}{n+1} + \cos \frac{j\pi}{n+1} \lambda \frac{\sin j\pi}{n+1}$$

$$\lambda \frac{\sin (i-1)j\pi}{n+1} = \lambda \frac{\sin ij\pi}{n+1} \cos \frac{j\pi}{n+1} - \cos \frac{j\pi}{n+1} \lambda \frac{\sin j\pi}{n+1}$$

add the two gives

$$\cancel{\sin \frac{ij\pi}{n+1}} (2 \cos \frac{j\pi}{n+1}) = \lambda \cancel{\sin \frac{ij\pi}{n+1}}$$

$$\therefore \lambda = 2 \cos \frac{j\pi}{n+1}$$

$$\therefore A \underline{x}^j = 2 \cos \frac{j\pi}{n+1} \underline{x}^j$$

3.90 we can write $B = [e, d, e]$ (note: $[x, x, x]$ means tri-diag form with elements e on super & subdiag & elements d on main diag)

$$B = dI + e[1, 0, 1]$$

by 3.91

$$B \underline{x} = \mu \underline{x} \quad \text{has eigenvalues } \mu = e\lambda + d$$

$$\text{where } \lambda = 2 \cos \frac{j\pi}{n+1} \quad \text{solving } A \underline{x}^j = \lambda^j \underline{x} \quad A: [1, 0, 1]$$

$$\therefore \mu = d + 2e \cos \frac{j\pi}{n+1}$$

3.95 Using power method to estimate spectral radius and corresponding eigenvector, for 20×20 matrix $A: [-1, 4, -1]$ and compare with exact results from 3.94

$$\lambda_{\max} = 4 - 2 \cos \frac{(20)\pi}{21} = 4 - 2 \cos(171.5^\circ) = 4 - 2(-.989) = 5.978$$

mod
16

B^m

	0	1	2	3	4	5	6	7
1	3	2	10	37	150	654	3012	14475
2	2	3	3	-2	-54	-396	-2427	-13838
3		4	4	9	80	189	1138	7145
4				8	15	14	-166	-1888
5					16	33	86	447
6						32	63	102
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m

λ

	1	2	3	4	5	6	7
1	3.333	3.7	4.4	4.84	~4.56	~4.67	~4.67
2	1.5	-1.67	27	~8	6.4	~5.6	~5.6
3	2	2.5	3.33	6.3	6.0	6.26	6.26
4		2	18.75	1.4	-12	~11.3	~11.3
5			2	2.4	~2.63	5.14	5.14
6				2	~2	~1.67	~1.67
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94							
95							
96							
97							
98							
99							
100							

etcetera

the eigenvector is obtained by taking eigenvalue obtained

& forming $(\frac{1}{\lambda} B)^m \approx y_1$

after fifty iterations using an average $\lambda_{\max} = \sum_{i=1}^{20} \lambda_i / 20$

$\lambda_{\max} = 5.9402219$, closest to this average value

where entries used are 6th or 15th

$$Bz^{(m)} = \begin{bmatrix} 1.0 \\ -10.63 \\ 13.70 \\ -14.76 \\ 14.02 \\ -11.99 \\ 9.29 \\ -6.43 \\ 3.72 \\ -1.21 \\ -1.21 \\ 3.72 \\ -6.43 \\ 9.29 \\ -11.99 \\ 14.02 \\ -14.76 \\ 13.70 \\ -10.63 \\ 5.82 \end{bmatrix}$$

since $By_i^* = \lambda y_i^* \sim Bz^{(m)}$

then $y_i^* = \frac{B}{\lambda} y_i^* =$

$$\begin{bmatrix} .168 \\ -1.79 \\ 2.31 \\ -2.49 \\ 2.36 \\ -2.02 \\ 1.56 \\ -1.08 \\ 6.26 \\ -2.04 \\ -2.04 \\ 6.26 \\ -1.08 \\ 1.56 \\ -2.02 \\ 2.36 \\ -2.49 \\ 2.31 \\ -1.79 \\ +.168 \end{bmatrix}$$

$$1. \quad P_0(x) = 1 \quad P_1(x) = x \quad P_2(x) = \frac{3}{2} \left[x^2 - \frac{1}{3} \right]$$

$$S_0 = 2 \quad S_1 = \frac{2}{3} \quad S_2 = \frac{2}{5}$$

$$f(x) = \sin \pi x$$

$$\langle f, P_0 \rangle = \int_{-1}^1 \sin \pi x \, dx = -\frac{1}{\pi} \cos \pi x \Big|_{-1}^1 = 0 \quad \Rightarrow d_0 = 0$$

$$\langle f, P_1 \rangle = \int_{-1}^1 x \sin \pi x \, dx = -\frac{x}{\pi} \cos \pi x \Big|_{-1}^1 + \frac{1}{\pi^2} \sin \pi x \Big|_{-1}^1 = \frac{2}{\pi} \quad \Rightarrow d_1 = \frac{2/\pi}{2/3} = .95493$$

$$\langle f, P_2 \rangle = \int_{-1}^1 \frac{3}{2} \left(x^2 - \frac{1}{3} \right) \sin \pi x \, dx = +\frac{1}{2\pi} \cos \pi x \Big|_{-1}^1 + \frac{3}{2} \left[-\frac{x^2}{\pi} \cos \pi x + \frac{2x}{\pi^2} \sin \pi x + \frac{2}{\pi^3} \cos \pi x \right]_{-1}^1 = 0$$

$$\Rightarrow d_2 = 0$$

$$\hat{f}(x) = d_0 + d_1 P_1(x) + d_2 P_2(x) = 0 + .95493 x + 0 \cdot \frac{3}{2} \left(x^2 - \frac{1}{3} \right) = .95493 x$$