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CSE 2201: Numerical Methods

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* "Mostly Mathematical. But algo related"

* Books: ① Balagurusamy - Numerical Methods

(Preferably
Bought)

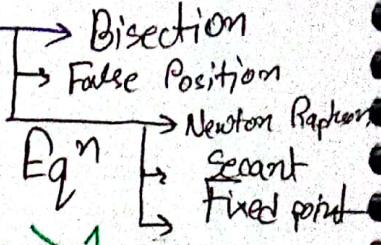
G. Shanker Rao - Numerical Analysis

* Quizzes: ② from Moodle (1.2 PTSD yay)

GOOD LUCK

Bisection Method (BG Ch - 6)

* Used to find roots of Non-Linear Eqⁿ



Find a root of $x^2 - 4x - 10 = 0$

$$\therefore x_{\text{root}} = \sqrt{\left(\frac{-4}{1}\right)^2 - 2\left(\frac{-10}{1}\right)} = 6$$

∴ The root is between -6 and 6

x	-6	-5	-4	-3	-2	-1	0	1	2	3	4	5	∞
$f(x)$	50	35	22	11	2	-4	-10	-13	-14	-13	-10	-5	2
					↓	↓	has one root	→ intersected	←	has another root			

Let us take $x_1 = -2$ and $x_2 = -1$

$$\therefore x_0 = \frac{-2 - 1}{2} = -1.5$$

$$\therefore f(-1.5) = \dots$$

* Take opposite sides in interval until they are roughly similar.

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Num Meth Lab

Monks

Attendance: 20%.

Online + Viva: 60% (5 online)

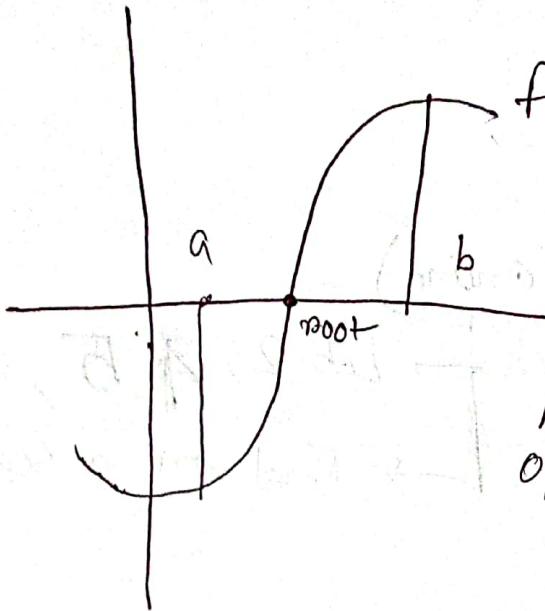
Final : 20%.

→ Lab 2, 4, 5, 7

→ Final on a separate date

GOOD LUCK

Bisection Method



The root of $f(x)$

$f(x)$ exists iff

$$f(a) * f(b) \leq 0$$

a and b
has to be
Opposite signs

if =,
a or b
will be root

Two types of methods: i) Bracketing \rightarrow fixed range
ii) Open and

*Initial Guess: Check if root exists int between a, b

*Wrong Guess: Find $\frac{a+b}{2}$, update guess based
on $\frac{a+b}{2}$. \rightarrow for c = +ve, update b
c = -ve, " a

Stop at,
 $f_{abs}(a-b) \rightarrow 0.01$

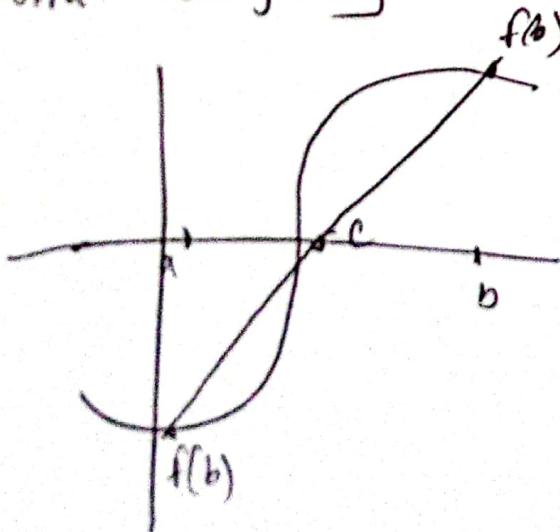
Loop continuation
Stopping Criteria:

1. $f(\text{root}) \geq 0.001$
2. $f(a) f_{abs}(a-b) \geq 0.001$
3. $f_{abs}((a-b)/a) \geq 0.001$

Check
Book
(G.4)

G.10 : Descartes
Rule of Sign

False position: Instead of middle point, find the point when joining $f(a)$ and $f(b) \rightarrow$ more accurate and saves than bisection



$$\text{root} = \frac{af(b) - bf(a)}{f(b) - f(a)}$$

Roots of Non-Linear Eqⁿ

* Non Linear Polynomial

* Transcendental Equations: (constant)^x

$$\hookrightarrow \frac{d}{dx}(a^x) = a^x \ln a$$

Search Bracket: - Range for which all roots are available
 - Largest possible roots.

For any polynomial:

$$a_n x^n + a_{n-1} x^{n-1} + a_{n-2} x^{n-2} + \dots + a_1 x + a_0 = 0,$$

$$\text{Largest possible root} = -\frac{a_{n-1}}{a_n}$$

$$\text{Search Bracket} = \sqrt{\left(\frac{a_{n-1}}{a_n}\right)^2 - 2\left(\frac{a_{n-2}}{a_n}\right)}$$

Stopping Criteria: [Check Book]

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Bisection Method: // Algo from Book // Convergence "

```

x0 = (x1 + x2) / 2;
if (f(x0) == 0)
{
    cout << x0 << endl;
    break;
}

```

* If nothing is mentioned → do only 4 iterations

- * For False position → " " 3 "
- * .. Newton-Raphson → " " 6 "
- * Secant → " " 4 "

False Position:

$$\text{replace } x_0 = \frac{x_1 + x_2}{2}$$

$$\text{with } x_0 = \frac{f(x_1)(x_2 - x_1)}{f(x_2) - f(x_1)}$$

* degree of convergence is high (more than Bisection)

Derive the formula for Bisection/False position Method

Newton Raphson - (6.8)

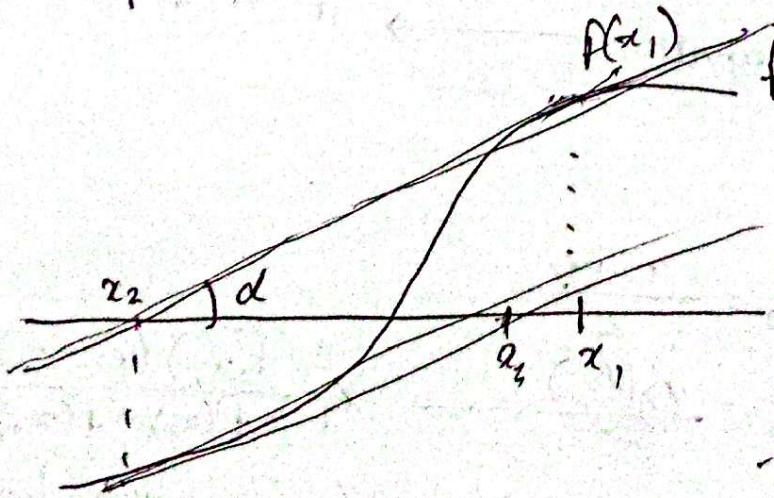
* Open end ~~has a~~ is uncontrollable

* Convergence NOT guaranteed

Check ⑧ Example 6.6

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

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



Hence,

$$\text{tangal} = \frac{f(x)}{x_1 - x_2}$$

$$\text{on, } f'(x) = \frac{f(x)}{x_1 - x_2}$$

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

Formulae
Derivation

Q. Solve using Newton-Raphson with initial value of 0
 $x^2 - 3x + 2 = 0$ viscosity

$$\therefore f(x) = x^2 - 3x + 2$$

$$\therefore f'(x) = 2x - 3$$

$$\therefore x_1 = x_0 - \frac{f(x_0)}{f'(x_0)} = 0 - \frac{f(0)}{f'(0)}$$

$$= 0.667$$

$$\therefore x_2 = 0.667 - \frac{f(0.667)}{f'(0.667)} = 0.9959$$

Q.

$$\therefore x_0 = 0.99 - \frac{0.001}{-1.0002} = 1.$$

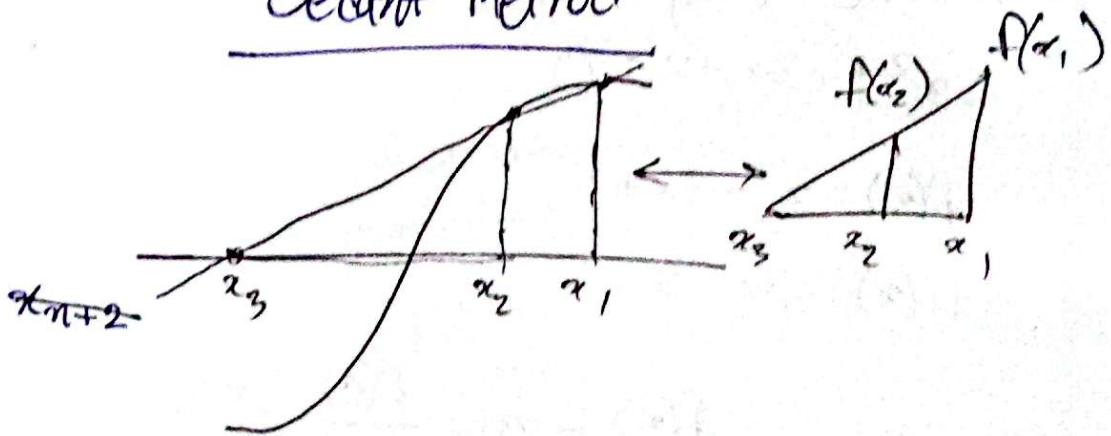
$$\therefore f(x_0) = f(1) = 0,$$

root = 1 (Ans)

+ If it ~~does not converge~~ OR $f'(x) = 0$,
 the answer is that it will not have an answer in that
 initial value

Limitations/Weaknesses of Newton-Raphson

Secant Method



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



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

Comparison between Newton Raphson and Secant

N - R

- Takes tangent to find x_2
- Probability of convergence is lower

Sec

- takes secant (using x_2 and x_1) to find x_3
- Probability of convergence is higher

All Possible Roots by Newton Method

Algo (in my own way): → divide the polynomial
with $x - x_n$ until
 $n = \text{degree of } eq^n$ all values are out

find root 1 ^{2nd}
while $n > 1 \rightarrow$, Last root

find root (using Newton Raphson method)

root x_{nr}

use synthetic division to make deflate

eq^n

$n - -$

$$\text{Root} = \frac{-a_0}{a_1}$$

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Synthetic Division: → Formula derivation

$$\frac{a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0}{x - x_n} = b_n x^n + b_{n-1} x^{n-1} + \dots + b_1 x + b_0; [b_n = 0]$$

\uparrow
 $g(x)$

$b_m = 0$
 $b_{i-1} = a_i + x_n b_i$

$\cancel{b_{i-1}}$

Good Luck

6.15

Hence,

$$x_m = 3$$

$$a_3 = 1, a_2 = -7, a_1 = 15, a_0 = -9$$

$$\therefore b_3 = 0$$

$$\begin{aligned}\therefore b_2 &= a_3 + b_3 x_m \\ &= 1\end{aligned}$$

$$\begin{aligned}\therefore b_1 &= a_2 + b_2 x_m \\ &= -7 + (1)(3) \\ &= -4\end{aligned}$$

$$\therefore b_0 = a_1 + b_1 x_m$$

$$= 3$$

$$\therefore \text{New eqn. } \equiv x^2 - 4x + 3 = 0$$

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* DeMorgan's Rule of Signs

* Horner's Rule → 6.5 Evaluation of Polynomial

$$\# 2x^3 - 4x^2 + 3x + 6 \text{ at } x = 2$$

$$\{(2x - 4)x + 3\}x + 6$$

Use P_n method
from book

At $x = 2$,

$$[(2(2) - 4)(2) + 3](2) + 6$$

$$= [(4 - 4)(2) + 3](2) + 6$$

$$= 3(2) + 6$$

$$= 12$$

Ch - 7Direct Solution of Linear Equations* 7.2. Existence of Solution :

- Unique solⁿ → intersecting two lines
- No " " → parallel [diff const]
- No unique / Infinite solⁿ → same eqⁿ
- * → ill conditioned (7.9) → almost same eqⁿ
 - ↳ check / identify
 - ↳ impact / effect **

- * Methods
- Gauss Elimination
 - " " with pivoting
 - " Jordan
 - triangular Factorization / LU decomposition
 - Matrix inversion

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#7.1. Solve using Gauss Elimination:

$$\begin{aligned} 3x + 2y + z &= 10 \\ 2x + 3y + 2z &= 14 \\ x + 2y + 3z &= 14 \end{aligned}$$

→ Step 1 : Make
Upper triangle
Matrix
by Forward
Elimination

→ Step 2 :
Back Substitution

#7.3. Solve using Gauss Elimination and Pivoting:

$2x_1$	$+ 2x_2 + x_3 = 6$	swap highest magnitude at the top
$4x_1$	$+ 2x_2 + 3x_3 = 4$	
x_1	$+ x_2 + 2x_3 = 0$	

* Gauss Jordan \rightarrow already done \rightarrow Example 7.4
 [check book]

LU decomposition / Triangular Factorization:

coefficient
variable { matrix

$$AX = B$$

divide into two
matrices L and U

$$LUX = B \rightarrow \text{Let,}$$

$$\begin{aligned} UX &= Z \\ LZ &= B \rightarrow \text{solve for } Z \end{aligned}$$

$$UX = Z \rightarrow \text{solve for } X$$

$$\begin{bmatrix} 1 & 0 & 0 \\ b_{21} & 1 & 0 \\ b_{31} & b_{32} & 1 \end{bmatrix} \times \begin{bmatrix} u_{11} & u_{12} & u_{13} \\ 0 & u_{22} & u_{23} \\ 0 & 0 & u_{33} \end{bmatrix} = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

~~ANSWER~~

$$L = \begin{bmatrix} 1 & 0 & 0 \\ \frac{2}{3} & 1 & 0 \\ \frac{1}{3} & \frac{4}{5} & 1 \end{bmatrix}, \quad U = \begin{bmatrix} 3 & 2 & 1 \\ 0 & \frac{5}{3} & \frac{4}{3} \\ 0 & 0 & \frac{24}{15} \end{bmatrix}$$

$$\therefore L_2 = \begin{bmatrix} 1 & 0 & 0 \\ \frac{2}{3} & 1 & 0 \\ \frac{1}{3} & \frac{4}{5} & 1 \end{bmatrix} \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix} = \begin{bmatrix} 10 \\ 14 \\ 14 \end{bmatrix}$$

$$\therefore z_1 = 10 \text{ cm}$$

$$\therefore \frac{2}{3} z_1 + z_2 = 14$$

$$\therefore z_2 = 14 - \frac{20}{3} = \frac{22}{3} \text{ (cm)}$$

$$\therefore \frac{1}{3} z_1 + \frac{4}{5} z_2 + z_3 = 14$$

$$\therefore z_3 = 14 - \frac{10}{3} - \frac{4 \times 22}{5 \times 3} = \cancel{\frac{21}{3}} - \cancel{\frac{88}{15}} - \frac{24}{5}$$

Now,

$$UX = Z$$

$$\text{or, } \begin{bmatrix} 3 & 2 & 1 \\ 0 & \frac{5}{3} & \frac{4}{3} \\ 0 & 0 & \frac{24}{15} \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 10 \\ \frac{22}{3} \\ \frac{272}{15} \end{bmatrix}$$

$$\therefore \frac{24}{15} z = \frac{72}{15}$$

$$\therefore z = 3 \text{ (Ans)}$$

$$\therefore \frac{5}{3}y + \frac{4}{3}z = \frac{22}{3}$$

$$\therefore z - y = \frac{3}{5} \times \left(\frac{22}{3} - \frac{4}{3} \times 3 \right)$$

$$\therefore y = 2 \text{ (Ans)}$$

$$\therefore 3x + 2y + z = 10$$

$$\therefore x = \frac{10 - 4 - 3}{3} = 1 \text{ (Ans)}$$

Cramer's Rule

$$x = \frac{D_x}{D}$$

$$y = \frac{D_y}{D}$$

$$z = \frac{D_z}{D}$$

$D_n = \begin{bmatrix} \text{const} \\ \text{const} \\ \text{const} \end{bmatrix}$

replace i^{th} column
with constant for D_i

Inversion Method

$$AX = B$$

or, $A^{-1}A X = A^{-1}B \rightarrow$ m left multiplication with A^{-1}

$$\text{or, } \boxed{X = A^{-1}B}$$

↳ left multiplication

≠
right multiplication

* Finding Inverse:

(i) For

$$\left[\begin{array}{ccc|ccc} a_{11} & a_{12} & a_{13} & 1 & 0 & 0 \\ a_{21} & a_{22} & a_{23} & 0 & 1 & 0 \\ a_{31} & a_{32} & a_{33} & 0 & 0 & 1 \end{array} \right],$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & a'_{11} & a'_{12} & a'_{13} \\ 0 & 1 & 0 & a'_{21} & a'_{22} & a'_{23} \\ 0 & 0 & 1 & a'_{31} & a'_{32} & a'_{33} \end{array} \right]$$

↙ A^{-1}

$$\therefore x_1 = a'_{11} b_1 + \dots$$

$$x_2 = - \dots$$

$$x_3 = \dots - \dots$$

$$(i.) A^{-1} = \frac{\text{Adj}(A)}{\text{Det}(A)}$$

$$A = \begin{bmatrix} 3 & 1 & 2 \\ 2 & -3 & -1 \\ 1 & 2 & 1 \end{bmatrix}$$

$$B = \begin{bmatrix} 5 \\ -3 \\ 4 \end{bmatrix}$$

$$\begin{aligned} \therefore |A| &= 3(-3+2) - 1(2+1) + 2(4+3) \\ &= -3 - 3 + 14 \\ &= 8 \end{aligned}$$

$\because |A| \neq 0$, A^{-1} exists

For cofactors matrix, of A^T

$$\begin{bmatrix} 3 & 2 & 1 \\ 1 & -3 & 2 \\ 2 & -1 & 1 \end{bmatrix}$$

$$\begin{array}{lcl} a_{11} = (-1)^{1+1} (-3+2) & a_{21} = (-1)^{2+1} (2+1) & a_{31} = (-1)^{3+1} (4+3) \\ = -1 & = -3 & = 7 \\ a_{12} = (-1)^{1+2} (1-4) & a_{22} = (-1)^{2+2} (3-2) & a_{32} = (-1)^{3+2} (6-1) \\ = -3 & = 1 & = -5 \\ a_{13} = (-1)^{1+3} (-1+8) & a_{23} = (-1)^{2+3} (-3-4) & a_{33} = (-1)^{3+3} (-9-2) \\ = 7 & = 7 & = -11 \end{array}$$

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$$\therefore \text{Adj}(A) = \begin{bmatrix} -1 & 3 & 2 \\ 3 & 1 & 7 \\ 7 & -5 & -11 \end{bmatrix}$$

$$\therefore A^{-1} = \frac{1}{8} \begin{bmatrix} -1 & 3 & 5 \\ 3 & 1 & 7 \\ 7 & -5 & -11 \end{bmatrix}$$

GOOD LUCK!

Ch - 8 : Iterative Methods

→ Jacobi Iteration Method

→ Gauss Seidel Method
↳ check book

Jacobi Iteration:

$$5x + 2y + z = 12$$

$$x + 4y + 2z = 15$$

$$x + 2y + 5z = 20$$

$$\therefore x = \frac{12 - 2y - z}{5}$$

$$\therefore y = \frac{15 - x - 2z}{4}$$

$$\therefore z = \frac{20 - x - 2y}{5}$$

↳ Convergence

Rule x

$$x^{i+1} = \frac{\dots - y^i - z^i}{\dots}$$

$$y^{i+1} = \frac{\dots + x^i + z^i}{\dots}$$

$$z^{i+1} = \frac{\dots + y^i + x^i}{\dots}$$

Ch - 10 : Regression

Curve Fit.

→ Curve Fitting: Regression
↳ Not Exact fitting

→ Curve Fitting: Interpolation
↳ touches all point

* We minimize SSE

Linear Reg.

Curve Fitting Regression

→ Straight Line

→ Power Function

→ Polynomial

→ Multiple Linear Regression

Derive
the express
for a
and b

For

for $y = a + bx$ and $y = ax^b \rightarrow$ we $\ln x$ and $\ln y$

$$b = \frac{n \sum x_i y_i - \sum x_i \sum y_i}{n \sum x_i^2 - (\sum x_i)^2} ; \quad n = \text{no. of points/obs}$$

observation

\uparrow replace x_i with $\ln x_i$
 y_i " $\ln y_i$

$$a = \frac{\sum y_i}{n} - b \frac{\sum x_i}{n} = \bar{y} - b \bar{x}$$

$$SSP = \sum (\epsilon)^2 = \sum (y_i - a - bx_i)^2$$

[Honesty,

Just check book
at this point]

Curve Fitting:

Interpolation

→ Equidistant points → $x_1 - x_0 = h$
 $x_2 - x_1 = h$
 $x_3 - x_2 = h$

+ Finite difference operators: Δ , ∇ , ∇^2

forward " " ; Δ

backgrounded " "

: S

~~central~~ " "

for

$$\begin{aligned}
 & \left\{ x_0, y_0 = f(x_0) \right\} \rightarrow \Delta y_0 = y_1 - y_0 \\
 & \left\{ x_1, y_1 = f(x_1) \right\} \rightarrow \Delta y_1 = y_2 - y_1 \\
 & \left\{ x_2, y_2 = f(x_2) \right\} \rightarrow \Delta y_2 = y_0 - y_1
 \end{aligned}$$

$$\Delta y_0 = y_1 - y_0$$

$$= \Delta f(x_0) = f(x_1) - f(x_0)$$

$$= f(x_0 + h) - f(x_0)$$

$$= \Delta f_0 = f_1 - f_0$$

$$\therefore \nabla y_0 = y_1 - y_0$$

$$\therefore \nabla y = y_3 - y_2$$

$$\nabla^2 y_3 = \Delta y_3 - \nabla y_2$$

$$\nabla^3 y_3 = \nabla^2 y_3 - \nabla^2 y_2$$

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Newton (P-Gregory's) Forward Interpolation Formula

Backward

Central

↓
some Equal Interval

Unequal Interval

Newton Divided Difference Interpolation Formula

Lagrange Interpolation

For inverse interpolation \rightarrow Lagrange.

* Check Interpolation from doc and then read from
Shankar Rao

* Ocean's Razon Hypothesis

Newton's Forward Interpolation:

* Multiple "Eq" can pass through same points

* Interpolation = Exact Fit

Error = 0

$$\begin{aligned} y_x &= \\ \phi(x) &= y_0 + u \cdot \frac{\Delta y_0}{1!} + \left[\frac{u(u-1)}{2!} \cdot \Delta^2 y_0 + \frac{u(u-1)(u-2)}{3!} \cdot \Delta^3 y_0 \right] \\ &\quad + \dots + \text{value to find} \\ u &= \frac{x - x_0}{h} \end{aligned}$$

x_0	$y =$	Δy	$\Delta^2 y$	$\Delta^3 y$
x_0	y_0	Δy_0	$\Delta^2 y_0$	$\Delta^3 y_0$
x_1	y_1	Δy_1	$\Delta^2 y_1$	
x_2	y_2	Δy_2	$\Delta^2 y_2$	
x_3	y_3			

Make sure n is constant

$$\# y_0 = 111.8034$$

$$\Delta y_0 = 0.0447$$

$$\Delta^2 y_0 = 0$$

$$u = 1.6$$

$$\begin{aligned} \therefore y_{(1)} &= 111.8034 + (1.6)(0.0447) \cdot 1 \\ &= 111.87492 \end{aligned}$$

~~Find~~ 101 \rightarrow Shanks

$$\therefore u = \frac{0.05 - 0.00}{0.1} = 0.5$$

$y_0 = 1$	z	y	Δy	$\Delta^2 y$	$\Delta^3 y$	$\Delta^4 y$
$\Delta y_0 =$	0	1	0.2214			
	0.1	1.2214	0.2704	0.049	0.0109	0.0023
	0.2	1.4918	0.3303	0.0599	0.0152	
	0.3	1.8221	0.4034	0.0731		
	0.4	2.225				

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Now,

$$y_{0.5} = y_0 + \frac{u}{1!} \cdot \Delta y_0 + \frac{u(u-1)}{2!} \cdot \frac{\Delta^2 y_0}{\Delta x} + \frac{u(u-1)(u-2)}{3!} \cdot \frac{\Delta^3 y_0}{\Delta x^2}$$

$$+ \frac{u(u-1)(u-2)(u-3)}{4!} \cdot \Delta^4 y_0$$

$$= 1 + (0.5)(2.2214) + \frac{(0.5)(-0.5)}{2} (0.049)$$

$$+ \frac{(0.5)(-0.5)(-1.5)}{6} (0.0109) + \frac{(0.5)(0.5)(-1.5)(-2.5)}{24} (0.0029)$$

$$\approx 2.105$$

(check later)

* If $\phi(x) \rightarrow x$ lies near first values of the table, we use Newton Forward Value

* If it lies near last values, we use Newton Backward

* If it lies near center: Gauss Forward
,, Backward

Stirling
Bessel's Everett $\Rightarrow X$

Newton's Backward Interpolation:

$$\phi(x) = y_n = y_n + \frac{u \nabla y_n}{1!} + \frac{u(u+1)}{2!} \nabla^2 y_n + \frac{u(u+1)(u+2)}{3!} \nabla^3 y_n + \dots$$

where,

$$u = \frac{x - x_n}{h}$$

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Gauss Forward Central Interpolation:

Use S

~~f(x)~~

$$\begin{aligned}
 f(x) = & y_0 + \frac{u}{1!} \Delta y_0 + \frac{u(u-1)}{2!} \Delta^2 y_{-1} + \frac{(u+1)u(u-1)}{3!} \Delta^3 y_{-2} \\
 & + \frac{\cancel{(u+2)}(u+1)u(u-1)(u-2)}{4!} \Delta^4 y_{-3} + \frac{(u+2)(u+1)u(u-1)(u-2)}{5!} \Delta^5 y_{-4}
 \end{aligned}$$

~~Answers~~

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Gauss Backward:

$$y = f(x) = y_0 + \frac{u}{1!} \Delta y_{-1} + \frac{u(u+1)}{2!} \Delta^2 y_{-1} + \frac{(u-1)u(u+1)}{3!} \Delta^3 y_{-2} \\ + \frac{u(u-1)u(u+1)(u+2)}{4!} \Delta^4 y_{-2} + \frac{(u-2)(u-1)u(u+1)(u+2)}{5!} \Delta^5 y_{-3}$$

Starting:

$$\frac{\text{Gauss Forward} + \text{Gauss Backward}}{2} = f(x)$$

Quiz - 3: Ch - 7, 8 → Check Matrix Inversion + Cramer's Rule } from sheet
Regression (Ch - 10)

→ Next Tuesday (27/2/25)

Finite Differences: Interpolation with ~~diff~~ unequal Interval

Newton's Divided Differences:

→ Newton Divided

→ Lagrange's

$$\Delta = \frac{y_1 - y_0}{x_1 - x_0}$$

$$\Delta^2 = \frac{\Delta y_1 - \Delta y_0}{x_2 - x_0}$$

$$\Delta^3 = \frac{\Delta^2 y_1 - \Delta^2 y_0}{x_3 - x_0}$$

α	$y f(\alpha)$	$\Delta f(\alpha)$	$\Delta^2 f(\alpha)$	$\Delta^3 f(\alpha)$	$\Delta^4 f(\alpha)$
5	150	121	$f[x_0, x_1]$	$f[x_0, x_1, x_2]$	$f[x_0, x_1, x_2, x_3]$
7	392	265	24	$f[x_1, x_2]$	$f[x_2, x_3]$
11	1452	457	32	1	
13	2306	917	46		
21	9702				

$$\begin{aligned}
 y = f(x) &= f[x_0] + (x - x_0) f[x_0, x_1] \rightarrow \Delta^1 \\
 &+ (x - x_0)(x - x_1) f[x_0, x_1, x_2] \rightarrow \Delta^2 \\
 &+ (x - x_0)(x - x_1)(x - x_2) f[x_0, x_1, x_2, x_3] \rightarrow \Delta^3
 \end{aligned}$$

Lagrange's :

$$\begin{aligned}
 y = f(x) &= \frac{(x - x_0)(x - x_1) \dots (x - x_n)}{(x_0 - x_1)(x_0 - x_2) \dots (x_0 - x_n)} y_0 \\
 &+ \frac{(x - x_0)(x - x_1) \dots (x - x_n)}{(x_1 - x_0)(x_1 - x_2) \dots (x_1 - x_n)} y_1 \\
 &+ \frac{(x - x_0)(x - x_1)(x - x_2) \dots (x - x_n)}{(x_2 - x_0)(x_2 - x_1)(x_2 - x_3) \dots (x_2 - x_n)} y_2
 \end{aligned}$$

B) Inverse Interpolation:

y is given \downarrow using y to find x
 $x = ?$

Replace x with y in Lagrange's formula :

$$x = \frac{(y-y_0)(y-y_1)(y-y_2)\dots(y-y_n)}{(y_0-y_1)(y_0-y_2)(y_0-y_3)\dots(y_0-y_n)} x_0$$

$$+ \frac{(y-y_0)(y-y_1)(y-y_2)\dots(y-y_n)}{(y_1-y_0)(y_1-y_2)(y_1-y_3)\dots(y_1-y_n)} x_1$$

+ . . . -

E : Shift Operator | S : Central Difference Operator

$$E y_0 = y_1 \quad \boxed{\mu \text{ = Averaging Operation}}$$

$$E y_i = y_{i+1} \quad | \quad E' y_i = y_{i+n}$$

$$\Delta y_i = y_{i+1} - y_i$$

$$\nabla y_i = y_i - y_{i-1}$$

$$S y_i = y_{i+\frac{1}{2}} - y_{i-\frac{1}{2}} ; i = \frac{p}{q}$$

$$\mu y_i = \frac{1}{2} \left[y_{i+\frac{1}{2}} + y_{i-\frac{1}{2}} \right] ; i = \frac{p}{q}$$

SR \rightarrow 4.1, 4.2, 4.3 \rightarrow 97 pg

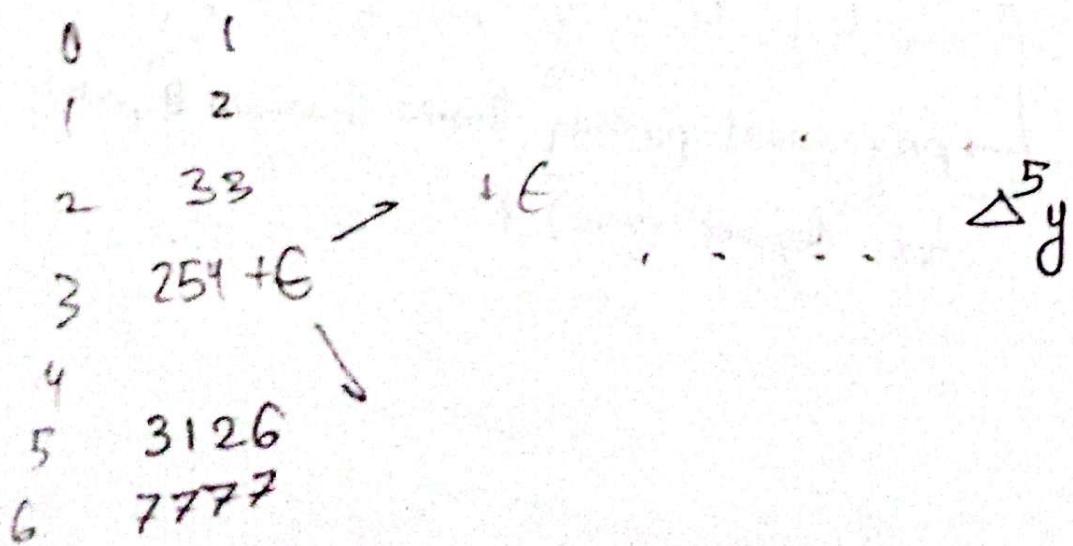
↳ polynomial passing through n points has degree $(n-1)$

* Find the missing value (4.1, 4.2, 4.3)

↳ the error in the table ($3^{24} - 8^3$)

There will be one
error, we use the
rest

#	-	0	1	2	3	4	5/6
							$y + \epsilon$



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Quiz #4

→ derivation
→ meth

- Newton Forward/Backward
- General Forward/Backward
- Stirling
- Newton Divided
- Lagrange
- Relation between operations

Next Week Tuesday 18/02/25

GOOD LUCK

Numerical Differentiation

Ch - 11

Forward Difference Quotient, $f'(x) = \frac{f(x+h) - f(x)}{h}$

Backward " "

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

Central " "

$$f'(x) = \frac{f(x+h) - f(x-h)}{2h}$$

use this if nothing
is mentioned

only when
function given

e.g. $f(x) = x^2$

Ex - 11.1, 11.2, 11.3 → Skip errors analysis

11.4

11.5, 11.6 → tabular

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* 2nd Order Derivative :

$$f''(x) = \frac{f(x+h) - 2f(x) + f(x-h)}{h^2}$$

* for a known function, we can also find
the errors \rightarrow find using formula
find using hand/heat

For tabulation ones check book for formula

Numerical Integration - Ch-9

* Derive the General Quadrature Formula:

* $\int_a^b f(x) dx \approx h \frac{f(x_0) + 2f(x_1) + f(x_2)}{3}$ " Trapezoidal Rule from GGF

↳ any num of interval (at least 6)

Simpson's 1/3rd Rule " "

↳ even numbers of interval

3/8th rule " "

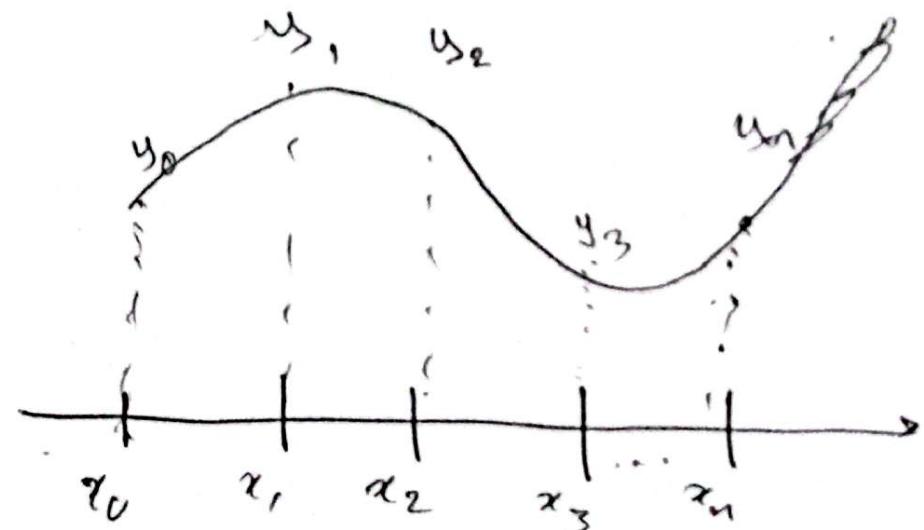
↳ at least multiple of 3

Weddle's rule " "

↳ multiple of 6

E

done



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$$\text{Trapezoidal} \Rightarrow I = \frac{h}{2} \left[(y_0 + y_n) + 2(y_1 + y_2 + y_3 + \dots + y_{n-1}) \right]$$

$$\text{for } n=6, I = \frac{h}{2} \left[(y_0 + y_6) + 2(y_1 + y_2 + y_3 + y_4 + y_5) \right]$$

$$\text{Simpson's 1/3rd} \Rightarrow I = \frac{h}{3} \left[(y_0 + y_n) + 4(y_1 + y_3 + y_5 + \dots) + 2(y_2 + y_4 + y_6) \right]$$

$$\text{for } n=6, I = \frac{h}{3} \left[(y_0 + y_6) + 4(y_1 + y_3 + y_5) + 2(y_2 + y_4) \right]$$

$$\text{Simpson's 3/8th} \Rightarrow I = \frac{3h}{8} \left[(y_0 + y_n) + 3(y_1 + y_2 + y_4 + y_5 + \dots) + 2(y_3 + y_6 + y_7) \right]$$

$$\text{for } n=6, I = \frac{3h}{8} \left[(y_0 + y_6) + 3(y_1 + y_2 + y_4 + y_5) + 2y_3 \right]$$

$$\text{Weddle's Rule} \Rightarrow I = \frac{3h}{10} \left[(y_0 + y_n) + (y_2 + y_4 + y_6 + y_8 + y_{10} + y_{12} + y_{14}) + 5(y_1 + y_3 + y_5 + y_7 + y_9 + y_{11} + y_{13} + y_{15}) + 6(y_3 + y_5 + y_7 + y_9 + y_{11} + y_{13} + y_{15}) + 2(y_6 + y_8 + y_{10} + y_{12} + y_{14}) \right]$$

$$\text{for } n=6, I = \frac{3h}{10} \left[(y_0 + y_6) + 5(y_1 + y_5) + 6(y_3) \right]$$

Rombeng Integration:

$\rightarrow \text{find } \frac{h}{2}, h, \frac{h}{2}, \frac{h}{4}$

```

graph TD
    Root["find h/2, h, h/2, h/4"] --> I1[I(h)]
    Root --> I2[I(h/2)]
    Root --> I3[I(h/4)]
    I1 --> Merge1["merge"]
    I2 --> Merge1
    I3 --> Merge1
    Merge1 --> I4["I(h, h/2)"]
    Merge1 --> I5["I(h/2, h/4)"]
    I4 --> Bottom[I(h, h/2, h/4)]
    I5 --> Bottom
  
```

$$\text{I}\left(h, \frac{h}{2}\right)$$

$$I(x,y) = \frac{1}{3} \left[4I\left(\frac{h}{2}\right) - I(h) \right]$$

$$\underline{\text{Ch - 4}} \longrightarrow \begin{array}{l} 4.7 \\ 4.9 \pi \pi \end{array}$$

+ Boxes
↓
device + panel

Approx and Errors

A Taxonomy of Errors

* ~~Significant digits~~

* Significant Digits:

* All non-zeroes

* If 0 → (i) in between digits

(ii) Trailing zeroes : We found using calculation
3.50

(iii) Leading zeroes ARE NOT counted

$$0.12 \times 10^{-5} = 0.\underset{\text{not significant}}{000}12$$

^{not significant}
_{significant}

(iv) Integers ending with 0 : 4500.0

Not significant

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#Ex 4.1, 4.2, 4.4

Accuracy: No. of significant digits

Precision: No. of decimal points

	Precision	Accuracy
4.3201	10^{-4} on 4	5 sd
4.32	10^{-2} on 2	3 sd
3600		Not allowed
3600.00		6 sd bc that makes perfect sense
0.000009	High Prec	
+2345.6		
123456.7		High Acc

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Error:

Inherent Errors:

→ Data/Empirical Error:

- Instrument error

→ Conversion/Representation Errors

$$\# 0.1 \times 2 = 0.2 - 0$$

$$0.2 \times 2 = 0.4 - 0$$

$$0.4 \times 2 = 0.8 - 0$$

$$0.8 \times 2 = 1.6 - 1$$

$$0.6 \times 2 = 1.2 - 1$$

$$0.2 \times 2 = 0.4 - 0$$

$$\therefore 0.1 = 0.00011\ 0011\ 0011\ldots$$

Note that Symmetric Rounding errors = $\frac{1}{2}$ Chopping errors



4.4 (67)

$$\begin{array}{r} 752.6833 \\ - 07526 \quad 835 \\ \hline \end{array}$$

Truncation Error:

~~Error~~
Error from taking a part from an infinite
series

Modelling Errors:

Error from ignoring things to make any
models

Blunders:

Human error

Absolute Error:

$$|x_t - x_a|$$

Relative Error:

$$\left| \frac{x_t - x_a}{x_a} \right| \text{ or } \left| \frac{x_t - x_a}{x_t} \right|$$

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Machine Epsilon:

Error from capacity of machine

* prove that machine absolute - - -

"Good Luck"

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Error Propagation → most imp

$x_t \rightarrow$ true value of x

$x_a \rightarrow$ approximate value of x

$$\therefore x_t = x_a + e_x$$

$$y_t = y_a + e_y$$

$$\text{error of } x_t + y_t \leq |e_x| + |e_y| \quad * \text{check proofs}$$

$$x_t - y_t \leq |e_x| + |e_y|$$

$$x_t * y_t \leq |x_a y_a| \left(\frac{|e_x|}{x_a} + \frac{|e_y|}{y_a} \right)$$

$$x_t / y_t \leq \left| \frac{x_a}{y_a} \right| \left(\frac{|e_x|}{|x_a|} + \frac{|e_y|}{|y_a|} \right)$$

$$e_{n,xy} = |e_{n,x}| + |e_{n,y}|$$

$$e_{n,\frac{x}{y}} = |e_{n,x}| + |e_{n,y}|$$

$$e_{n,x+y} \leq \left| \frac{x_a}{x_a + y_a} \right| |e_{n,x}| + \left| \frac{y_a}{x_a + y_a} \right| |e_{n,y}|$$

Regression Analysis

(using Gradient Descent)

→ Derive the expression for:

↳ Linear regression
↑
↳ gradient descent algorithm

↳ Multiple Linear Regression

↳ polynomial regression with degree 2, 3
↳ using with variable conversion
↳ without " "

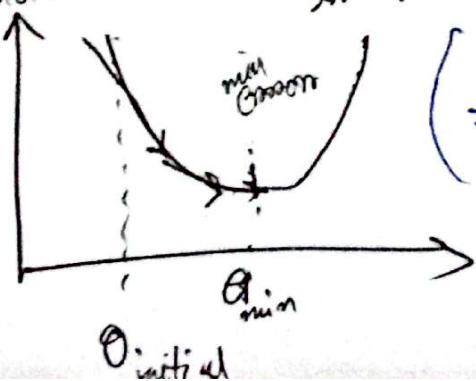
→ Math → Update $\theta_0, \theta_1, \dots$

learning rate / step size

$$\text{new } \theta_i = \text{old } \theta_i - \alpha \cdot \frac{\partial}{\partial \theta_i} (\text{cost function})$$

Cost function

↓
small



$$\left(\frac{\partial}{\partial \theta_0} (\text{cost}), \frac{\partial}{\partial \theta_1} (\text{cost}) \right) \rightarrow \text{Gradient}$$

$$\frac{1}{2m} \sum_{i=1}^m (\theta_0 + \theta_1 x_i - y)^2$$

↳ cost function

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Final:

Part 1:

- defⁿ
- deminuation * *
- descriptive

} Algo / short Q

Part 2:

- math / solve

Part 3:

- Regression Analysis

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* Ø Update :

Model: Linear $y = \theta_0 + \theta_1 x$.

$$\text{Mult Linear} \rightarrow \text{Ans} = \theta_0 + \theta_1 x_1 + \theta_2 x_2$$

$$3 \text{varm} = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \theta_3 x_3$$

Polynomial \rightarrow 2 degree $\therefore y = \theta_0 + \theta_1 x + \theta_2 x^2$

$$3 \text{ degree : } y = \theta + \theta_1 x + \theta_2 x^2 + \theta_3 x^3$$

Rule:

$$\text{new } \theta_i = \text{old } \theta_i - \alpha \frac{\partial}{\partial \theta_i} (\text{cost function})$$

$$\text{Cost function: } \frac{1}{2m} \sum_{i=1}^m (\text{error})^2$$