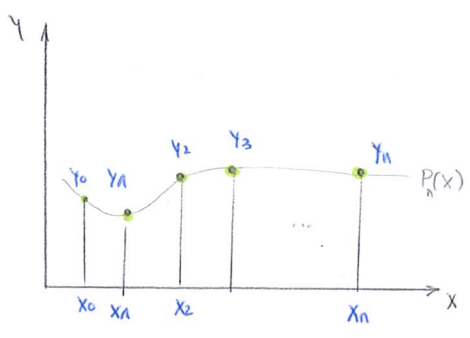


# CH. 1. INTERPOLATION

## LAGRANGE METHOD



\* Find  $P_n(x) \in \mathbb{P}_n / P_n(x_i) = y_i \quad (i=0, \dots, n)$

$$P_n(x) = C_0 + C_1 x + C_2 x^2 + \dots + C_n x^n \quad ((x_i \neq x_j)) \quad (n+1 \text{ DEGREES OF FREEDOM})$$

(LAGRANGE POLYNOMIAL)

$\left\{ \begin{array}{l} x_i \equiv \text{nodes} \\ (x_i, y_i) \equiv \text{interp. points / data} \end{array} \right.$

$$\begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix} \begin{pmatrix} C_0 \\ C_1 \\ \vdots \\ C_n \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_n \end{pmatrix}$$

(ALSO USED TO PROVE UNIQUENESS)

↳ NEVER SINGULAR (ALWAYS  $\neq 0$ )

### LAGRANGE INTERPOLATION

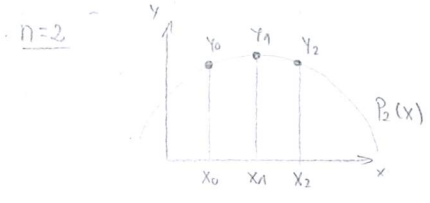
- Non-oscillating
- Non-piecewise

### EXISTENCE AND UNIQUENESS THEOREM (∃!)

\*  $\exists! P_n(x) \in \mathbb{P}_n / P_n(x_i) = y_i \quad (i=0, 1, \dots, n)$

PROOF:

#### a) EXISTENCE



$$P_2(x) = \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} \cdot y_0 + \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} \cdot y_1 + \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)} \cdot y_2 \in \mathbb{P}_2$$

$$P_n(x) = \sum_{i=0}^n \left( \prod_{\substack{j=0 \\ j \neq i}}^n \frac{(x-x_j)}{(x_i-x_j)} \right) \cdot y_i \quad (\text{THE LAGRANGE REPRESENTATION OF A POLYNOMIAL})$$

PRODUCT  $\rightarrow L_i(x)$  (LAGRANGE BASED FUNCTION)

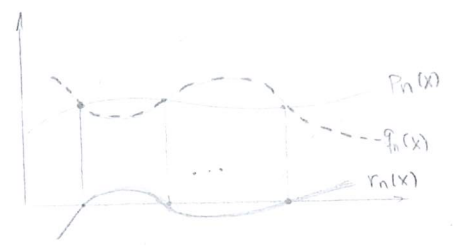
#### b) UNIQUENESS : reduction to absurdity

\* Assume  $P_n(x), Q_n(x)$  2 different solutions

$$r_n(x) = P_n(x) - Q_n(x) \in \mathbb{P}_n$$

with  $n+1$  roots against the fundamental th. of algebra

unless:  $r_n(x) = 0 \rightarrow Q_n(x) = P_n(x)$



• **EXERCISE 2**  $P_3(x)$  by  $(1,6), (2,4), (3,2), (4,6)$  (GAUSS VS. LAGRANGE)

a) via linear system (GAUSS) (of eqs.)

$$* P_3(x) = C_0 + C_1x + C_2x^2 + C_3x^3$$

$$\begin{pmatrix} 1 & 1^1 & 1^2 & 1^3 \\ 1 & 2^1 & 2^2 & 2^3 \\ 1 & 3^1 & 3^2 & 3^3 \\ 1 & 4^1 & 4^2 & 4^3 \end{pmatrix} \begin{pmatrix} C_0 \\ C_1 \\ C_2 \\ C_3 \end{pmatrix} = \begin{pmatrix} 6 \\ 4 \\ 2 \\ 6 \end{pmatrix}$$

→ AUGMENTED MATRIX

$$\begin{pmatrix} 1 & 1 & 1 & 1 & 6 \\ 1 & 2 & 4 & 8 & 4 \\ 1 & 3 & 9 & 27 & 2 \\ 1 & 4 & 16 & 64 & 6 \end{pmatrix} \begin{matrix} < R_2 - R_1 > \\ < R_3 - R_1 > \\ < R_4 - R_1 > \end{matrix} \begin{pmatrix} 1 & 1 & 1 & 1 & 6 \\ 0 & 1 & 3 & 7 & -2 \\ 0 & 2 & 8 & 26 & -4 \\ 0 & 3 & 15 & 63 & 0 \end{pmatrix} \begin{matrix} < R_3 - 2R_2 > \\ < R_4 - 3R_2 > \end{matrix} \begin{pmatrix} 1 & 1 & 1 & 1 & 6 \\ 0 & 1 & 3 & 7 & -2 \\ 0 & 0 & 2 & 12 & 0 \\ 0 & 0 & 6 & 42 & 6 \end{pmatrix} < R_4 - 3R_3 >$$

$$\begin{pmatrix} 1 & 1 & 1 & 1 & 6 \\ 0 & 1 & 3 & 7 & -2 \\ 0 & 0 & 2 & 12 & 0 \\ 0 & 0 & 0 & 6 & 6 \end{pmatrix}$$

$$\begin{cases} C_0 + C_1 + C_2 + C_3 = 6 \rightarrow C_0 = 2 \\ C_1 + 3C_2 + 7C_3 = -2 \rightarrow C_1 = 9 \\ 2C_2 + 12C_3 = 0 \rightarrow C_2 = -6 \\ 6C_3 = 6 \rightarrow C_3 = 1 \end{cases}$$

$$P_3(x) = 2 + 9x - 6x^2 + x^3$$

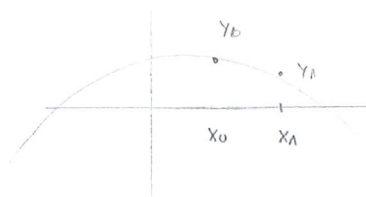
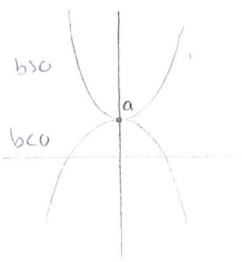
b) via Lagrange REPRESENTATION

$$* P_3(x) = \sum_{i=0}^3 \left( \prod_{\substack{j=0 \\ j \neq i}}^3 \frac{(x-x_j)}{(x_i-x_j)} \right) y_i$$

$$P_3(x) = \frac{(x-2)(x-3)(x-4)}{(1-2)(1-3)(1-4)} \cdot 6 + \frac{(x-1)(x-3)(x-4)}{(2-1)(2-3)(2-4)} \cdot 4 + \frac{(x-1)(x-2)(x-4)}{(3-1)(3-2)(3-4)} \cdot 2 + \frac{(x-1)(x-2)(x-3)}{(4-1)(4-2)(4-3)} \cdot 6$$

• **EXERCISE 1**

Study the existence and uniqueness of  $p(x) = a + bx^2$  by  $(x_0, y_0), (x_1, y_1)$   
PARABOLA



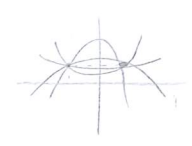
$$\begin{pmatrix} 1 & x_0^2 \\ 1 & x_1^2 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \end{pmatrix}$$

$$\begin{pmatrix} 1 & x_0^2 & | & y_0 \\ 1 & x_1^2 & | & y_1 \end{pmatrix} < R_2 - R_1 > = \begin{pmatrix} 1 & x_0^2 & | & y_0 \\ 0 & x_1^2 - x_0^2 & | & y_1 - y_0 \end{pmatrix}$$

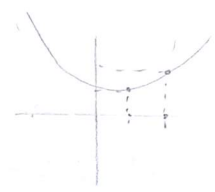
$$\begin{pmatrix} 1 & X_0^2 & | & Y_0 \\ 0 & X_1^2 - X_0^2 & | & Y_1 - Y_0 \end{pmatrix}$$

• CASE 1 :  $X_1^2 - X_0^2 = 0$

$\begin{cases} X_1 = X_0 \\ X_1 = -X_0 \end{cases} \begin{cases} Y_0 = Y_1 \rightarrow \infty \text{ many solutions} \\ \text{(compatible and determined system)} \\ Y_0 \neq Y_1 \rightarrow \text{NO SOLUTION} \end{cases}$



• CASE 2 :  $X_1^2 - X_0^2 \neq 0$  — system with ONLY ONE solution  
( $X_1 \neq \pm X_0$ ) (compatible system)

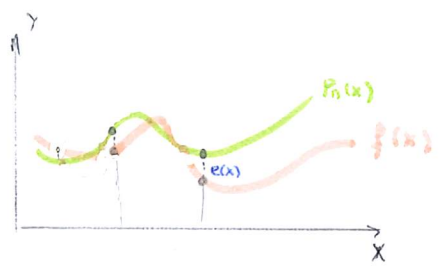


ERRORS

General def: error = exact - approx.

$\begin{cases} \text{appr} > \text{exact} \rightarrow \text{error} < 0 \rightarrow \text{ERROR IN EXCESS} \\ \text{appr} < \text{exact} \rightarrow \text{error} > 0 \rightarrow \text{ERROR IN DEFECT} \\ \text{(FOR DEFECT)} \end{cases}$

TRUNCATION ERRORS : errors with exact arithmetic



\*  $e(x) = P_n(x) - f(x)$

\* even though the arithmetic is exact we have an error, it will be because of the error of the method used.

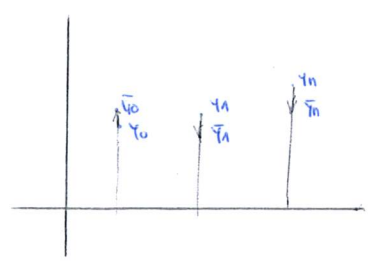
ROUNDING ERRORS

Ex: 3,1416 ...

02/02/21

ERROR PROPAGATION: AMPLIFICATION FACTOR (AF) AND INSTABILITY

Instead of  $y_i$  we have  $\bar{y}_i$  (perturbed values) "all else exact".



$E \geq |y_i - \bar{y}_i| \quad \forall i$

upper bound

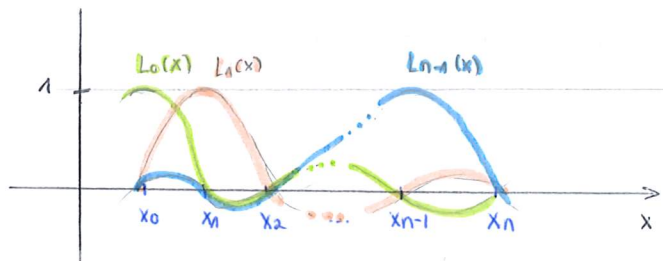
$$\begin{cases} p(x) = \sum_{i=0}^n L_i(x) \cdot y_i \\ \bar{p}(x) = \sum_{i=0}^n L_i(x) \cdot \bar{y}_i \end{cases}$$

$$e_d(x) = p(x) - \bar{p}(x) \rightarrow e_d(x) = \sum_{i=0}^n L_i(x) (y_i - \bar{y}_i)$$

\* applying absolute values:

$$|e_d(x)| = \left| \sum_{i=0}^n L_i(x) (y_i - \bar{y}_i) \right| \leq \sum_{i=0}^n |L_i(x)| |y_i - \bar{y}_i| \leq \epsilon \cdot \underbrace{\sum_{i=0}^n |L_i(x)|}_{\text{A.F.}}$$

$$* \text{AF} = \sum_{i=0}^n |L_i(x)| = \sum^+ L_i(x) + \sum^- |L_i(x)| = \sum^+ L_i(x) - \sum^- L_i(x) \quad (\text{Amplification Factor})$$



$$p_n(x) = \sum_{i=0}^n L_i(x) \cdot f(x_i) \equiv 1$$

$$\sum^+ L_i(x) + \sum^- L_i(x) \equiv 1$$

$$L_0(x) = \frac{(x-x_1) \dots (x-x_n)}{(x_0-x_1) \dots (x_0-x_n)} \rightarrow L_i(x) = \delta_{ij} \begin{cases} 0 & i \neq j \\ 1 & i = j \end{cases}$$

→ KRÖNCKER'S DELTA

so now we substitute:  $\sum^+ L_i(x) + \sum^- L_i(x) = 1$  in our A.F

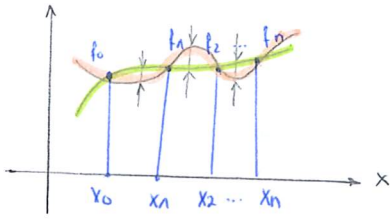
$$\text{AF} = \sum^+ L_i(x) - \sum^- L_i(x) = (1 - \sum^- L_i(x)) - \sum^- L_i(x) = 1 - 2 \sum^- L_i(x) = 1 + 2 \sum^- |L_i(x)|$$

↳ > 1 always

when  $\text{AF} \gg 1 \rightarrow \text{instability}$

FUNDAMENTAL THEOREM OF LAGRANGE TRUNCATION INTERPOLATION ERROR

ABOUT LAGRANGE INTERPOLATION NOT REPRESENTATION



$e(x) = f(x) - p_n(x)$  TRUNC. EMOR.

THEOREM:

$x_i, x \in [a, b]$   
 $f \in C^{(n+1)}([a, b])$   
 ↳ smoothness where  
 ↳ continuous and smooth

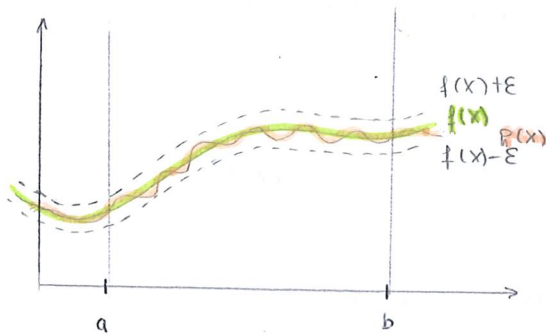
GREEK LETTER

$\exists \xi \in [a, b] / e(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \pi(x)$

- \*  $e(x_i) = 0 \forall (bc \pi(x_i) = 0)$
- \*  $f \in \mathbb{P}_n \Rightarrow e(x) = 0$

\*  $\pi(x) = (x-x_0)(x-x_1) \dots (x-x_n) = \prod_{i=0}^n (x-x_i)$

WEIERSTRASS' UNIFORM APPROXIMATION THEOREM



↳ UPPER BOUND (ga'ko muga)  
 \*  $M_{n+1} = \max_{x \in [a, b]} |f^{(n+1)}(x)|$

$|e(x)| = \left| \frac{f^{(n+1)}(\xi)}{(n+1)!} \pi(x) \right| \leq \frac{M_{n+1}}{(n+1)!} |\pi(x)| \leq \frac{M_{n+1}}{(n+1)!} (b-a)^{n+1}$

when  $n \uparrow$   $f^{(n+1)}$  can increase to  $\infty$  faster than  $(n+1)!$  (creo)  $\rightarrow$  no per meter mas va a ser mejor

THEOREM:  $f \in C^0([a, b]) \Rightarrow \forall \epsilon > 0, \exists p \in \mathbb{P} / |f(x) - p(x)| \leq \epsilon \forall x \in [a, b]$

uniform interpolation is possible thanks to Weierstrass

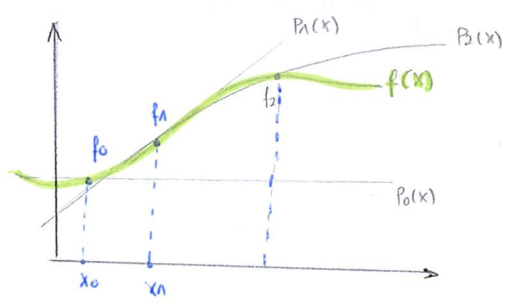
NON-PERMANENCE OF THE NODES

$$P_n(x) = \frac{(x-x_1)(x-x_2)\dots}{(x_0-x_1)\dots(x_0-x_n)} \cdot f_0 + \frac{(x-x_0)\dots}{(\dots)(\dots)} \cdot f_1 + \dots + \frac{(\dots)(\dots)}{(\dots)(\dots)} \cdot f_n$$

← If we add now a node  $n+1$ , all the terms will change

Is a disadvantage of Lagrange when we add a node, because we have to rewrite it all over again.

THE NEWTON REPRESENTATION OF LAGRANGE POLYNOMIAL



$$P_0(x) = f_0 = f(x_0) = a_0$$

$$P_1(x) = P_0(x) + h_1(x) = P_0(x) + a_1(x-x_0)$$

...

$$P_k(x) = P_{k-1}(x) + h_k(x) = P_{k-1}(x) + a_k(x-x_0)\dots(x-x_{k-1})$$

...

$$P_n(x) = P_{n-1}(x) + h_n(x) = a_0 + a_1(x-x_0) + \dots + a_n(x-x_0)(x-x_1)\dots(x-x_{n-1})$$

DEFINITION:  $h_k(x) = \underbrace{P_k(x)}_{\in \mathbb{P}_k} - \underbrace{P_{k-1}(x)}_{\in \mathbb{P}_k} \in \mathbb{P}_k$   $h_k(x_i) = 0 \quad (i=0, \dots, k-1)$

$\neq 0$  in general

\* whatever has to be added is a polynomial of degree  $k-1$  or less.

$h_k(x) \in \mathbb{P}_k \rightarrow h_k(x) = a_k(x-x_0)(x-x_1)\dots(x-x_{k-1})$

↳ LEADING COEFF

\* Leading coeff. of a polynomial = value of the coeff. ↗ that goes with the highest order.

$$P_n(x) = \sum_{i=0}^n a_i \prod_{j=0}^{i-1} (x-x_j)$$

$a_k =$  divided difference = leading coeff of  $P_k(x)$

DIVIDED DIFFERENCES

Notations:

↳  $a_k = f[x_0, x_1, \dots, x_k] = f_{k,k}$  (Depends on  $x_0, x_1, \dots, x_k \wedge f_0, f_1, \dots, f_k$ )

↳  $f_{i,k} = f[x_i, x_{i+1}, \dots, x_k]$

↳  $f_{i,0} = f[x_i] = f(x_i)$

## PROPERTIES

\* Permutation or permanence of the div. coeff.

$$* f_{i,k} = \frac{f_{i,k-1} - f_{i-1,k-1}}{x_i - x_{i-k}}$$

we use the formula to fill the table

## TABLE OF DIVIDED DIFFERENCE (TDD)

$x_i$	$f(x_i) = f_{i,0}$	$f_{i,1}$	$f_{i,2}$	...	$f_{i,n}$
$x_0$	$f_{0,0} = a_0$	—	—	...	—
$x_1$	$f_{1,0}$	$f_{1,1} = a_1$	—	...	—
$x_2$	$f_{2,0}$	$f_{2,1} = x$	$f_{2,2} = a_2$	...	—
$x_3$	$f_{3,0}$	$f_{3,1} = x$	$f_{3,2} = x$	...	—
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$x_{n-1}$	$f_{n-1,0}$	$f_{n-1,1} = x$	$f_{n-1,2} = x$	...	—
$x_n$	$f_{n,0}$	$f_{n,1} = x$	$f_{n,2} = x$	...	$f_{n,n} = a_n$

$a_0, a_1, \dots$  PRINCIPAL DIVIDED DIFFERENCES

## EXERCISE 6

a) TDD of  $f(x) = x^3$  on  $x_i = \{0, 1, 3, 4\} \rightarrow$  NEWTON RE.  $f_{i,k} = \frac{f_{i,k-1} - f_{i-1,k-1}}{x_i - x_{i-k}}$

$x_i$	$f_{i,0}$	$f_{i,1}$	$f_{i,2}$	$f_{i,3}$
0	0	—	—	—
1	1	1	—	—
3	27	13	4	—
4	64	37	8	1

FOR LAGR.

$$f_{1,1} = \frac{f_{1,0} - f_{0,0}}{x_1 - x_0} = \frac{1 - 0}{1 - 0} = 1$$

$$f_{2,1} = \frac{f_{2,0} - f_{1,0}}{x_2 - x_1} = \frac{27 - 1}{3 - 1} = 13$$

$$f_{3,1} = \frac{f_{3,0} - f_{2,0}}{x_3 - x_2} = \frac{64 - 27}{4 - 3} = 37$$

$$f_{3,2} = \frac{f_{3,1} - f_{2,1}}{x_3 - x_1} = \frac{37 - 13}{4 - 1} = 8$$

$$P_3(x) = 0 + 1(x-0) + 4(x-0)(x-1) + 1(x-0)(x-1)(x-3) = \dots = x^3$$

b) Lagrange form of the pol.

$$P_3(x) = \frac{(x-0)(x-3)(x-4)}{(1-0)(1-3)(1-4)} \cdot 1 + \frac{(x-1)(x-4)}{(0-1)(0-3)} \cdot 27 + \frac{(x-1)(x-3)}{(0-1)(0-4)} \cdot 64 = x^3$$

ALGORITHM AND CODE

$X$	$F$	$F$	$F$
$x_i$	$f_{i,0}$	$f_{i,1}$	$f_{i,2}$
x	x	x	x
x	x	□	□
x	x	□	△
⋮	⋮	⋮	⋮
x	x	□	△

$$F = \begin{pmatrix} f_0 \\ f_1 \\ \vdots \\ f_n \end{pmatrix} \rightarrow F = \begin{pmatrix} f_{0,0} = a_0 \\ f_{1,1} = a_1 \\ \vdots \\ \square \end{pmatrix} \rightarrow F = \begin{pmatrix} \end{pmatrix}$$

(OVERWRITE OF VALUES FOR F)

we start substituting from the bottom

$$\begin{cases} X = [x_0, x_1, \dots, x_n]^T \\ F = f(X) \end{cases}$$

PROGRAMMING THE CALC. OF F

```

X = [ x0, x1, x2, ..., xn ]';
F = [ f(x0), f(x1), f(x2), ..., f(xn) ]';

nn = numel(X); % (number of nodes)

n = nn - 1; % (the "n" of the "theory")

for k = 1:n
    for row = [nn:-1:(k+1)]-UNTL
        F(row) = (F(row) - F(row-1)) / (X(row) - X(row-k));
    end
end
end
    
```



HORNER'S ALGORITHM TO EVALUATE A POLYNOMIAL

$$P_4(x) = \underbrace{C_0}_0 + \underbrace{C_1 x}_1 + \underbrace{C_2 x^2}_2 + \underbrace{C_3 x^3}_3 + \underbrace{C_4 x^4}_4$$

↳ "DIRECTLY" → FLOPS: (14)

(FLOP → floating point operation)  
↳ ⊕ or ⊙

$$P_4(x) = \underbrace{C_0}_0 + x \left\{ \underbrace{C_1}_1 + x \left[ \underbrace{C_2}_2 + x \left( \underbrace{C_3}_3 + x \underbrace{C_4}_4 \right) \right] \right\}$$
 (nested product) → FLOPS: (8)

$$P_4(x) = C_0 + x \left\{ C_1 + x \left[ C_2 + x \left( C_3 + x C_4 \right) \right] \right\}$$

$\underbrace{\hspace{10em}}_{b_0}$   
 $\underbrace{\hspace{5em}}_{b_1}$   
 $\underbrace{\hspace{2em}}_{b_2}$   
 $\underbrace{\hspace{1em}}_{b_3}$   
 $\underbrace{\hspace{0.5em}}_{b_4}$

→ RUFFINI'S TABLE

	$C_4$	$C_3$	$C_2$	$C_1$	$C_0$
$x$		$x b_4$	$x b_3$	$x b_2$	$x b_1$
	$b_4$	$b_3$	$b_2$	$b_1$	$b_0 = P_4(x)$

HORNER-LIKE ALGORITHM (NESTED PRODUCTS) (OPTIMAL)

$$P_n(x) = a_0 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + \dots + a_{n-1}(x-x_0)(x-x_1)\dots(x-x_{n-2}) + a_n(x-x_0)\dots(x-x_{n-2})(x-x_{n-1})$$

$$P_n(x) = a_0 + (x-x_0) \left\{ a_1 + a_2(x-x_1) + \dots + a_{n-1}(x-x_1)\dots(x-x_{n-2}) + a_n(x-x_1)\dots(x-x_{n-2})(x-x_{n-1}) \right\}$$

$$P_n(x) = a_0 + (x-x_0) \left\{ a_1 + (x-x_1) \left[ a_2 + (x-x_2) \left( \dots \left( a_{n-2} + (x-x_{n-2}) \left( a_{n-1} + (x-x_{n-1}) a_n \right) \right) \right) \right] \right\}$$

$\underbrace{\hspace{10em}}_{b_0}$   
 $\underbrace{\hspace{5em}}_{b_1}$   
 $\underbrace{\hspace{2em}}_{b_2}$   
 $\dots$   
 $\underbrace{\hspace{1em}}_{b_{n-2}}$   
 $\underbrace{\hspace{1em}}_{b_{n-1}}$   
 $\underbrace{\hspace{0.5em}}_{b_n}$

\* PROGRAMMING \*

% X = nodes, F = principal div diffs

bi = F(end);

nn = length or numel (X); % no of nodes

for i = (nn-1):-1:1 EVALUATING POINT

bi = bi \* ((X - X(i)) + F(i));

end

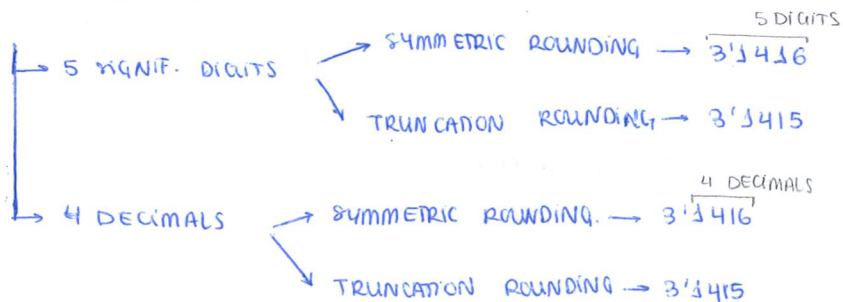
Pncf x = bi;

\*  $i \neq n$  → because  $n$  goes to 0 and in OCTAVE we can't  
 ↳ si queremos  $n = n-1$ ; for  $i = n-1:-1:0$  ... then  $(i+1)$   
 (MORE LITERAL)

# ROUNDING OF NUMBERS

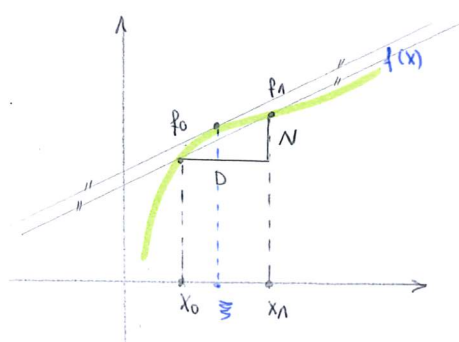
## SIGNIFICANT DIGITS VS. DECIMALS ⊕ SYMMETRIC AND TRUNCATION ROUNDING

(EX)  $\pi \rightarrow 3.141592654 \dots$



\* If we are said to imitate a computer  $\rightarrow$  we round <sup>in</sup> every step

## RELATION BETWEEN DIVIDED DIFFERENCES (DD) AND DERIVATIVES



$\begin{cases} N = \text{numerator} \\ D = \text{denominator} \end{cases}$

$$f[x_0, x_1] = \frac{f(x_1) - f(x_0)}{x_1 - x_0} = \frac{N}{D} = \text{slope}$$

$$* f \in C^1([x_0, x_1]) \Rightarrow \exists \xi \in [x_0, x_1] / f'(\xi) = \frac{f_1 - f_0}{x_1 - x_0}$$

$$* f \in C^2([x_0, x_1]) \Rightarrow \exists \xi \in [x_0, x_1] / f[x_0, x_1] = \frac{f'(\xi)}{(1)}$$

### THEOREM:

$$\left. \begin{array}{l} f \in C^n([a, b]) \\ x_0, x_1, \dots, x_n \in [a, b] \end{array} \right\} \Rightarrow \exists \xi \in [a, b] / f[x_0, x_1, \dots, x_n] = \frac{f^{(n)}(\xi)}{n!}$$

TABLES OF DIV. DIFF WITH A CONSTANT (ZERO) COLUMN

$f \in P_k$

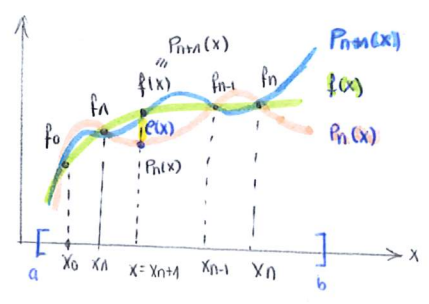
$x_i$	$f_{i,0}$	$f_{i,1}$	...	$f_{i,k}$	$f_{i,k+1}$
$x_0$	$f_{0,0}$	—	...	—	—
$x_1$	$f_{1,0}$	$f_{1,1}$	...	—	—
$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$	$\vdots$
$x_{n-1}$	$f_{n-1,0}$	$f_{n-1,1}$	...	C	0
$x_n$	$f_{n,0}$	$f_{n,1}$	...	C	0

$\downarrow$  KTE       $\downarrow$  0

$f \in P_k \Rightarrow$  col.  $f_{i,k} = kTE$   
 SUFFICIENT CONDITION

TRUNCATION ERROR BY LAGRANGE DIFF. EXPRESSED WITH D.D.

$$e(x) = f[x_0, x_1, \dots, x_n, x] \quad \Pi(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \Pi(x)$$



PROOF?

$$\begin{aligned}
 * e(x) &= f(x) - P_n(x) = P_{n+1}(x) - P_n(x) = h_{n+1}(x) = \\
 &= a_{n+1}(x-x_0)(x-x_1)\dots(x-x_n) = a_{n+1} \Pi(x) \\
 \hookrightarrow e(x) &= \underbrace{f[x_0, x_1, \dots, x_n, x]}_{\text{EXACT EQ.}} \underbrace{\Pi(x)}_{\text{DD OF ORDER } n+1}
 \end{aligned}$$

so if  $e(x) = f[x_0, x_1, \dots, x_n, x] \cdot \Pi(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \Pi(x)$   
 for some  $\xi \in [a, b] / x_i \in [a, b] \quad (i=0, \dots, n)$

best est. error = best est. exact - approx.

10/02/21

EXERCISE 10

$x_i$	0	0'2	0'3	0'4	0'7	0'9
$f(x_i)$	132'651	148'877	157'464	166'375	195'12	216

a) What is the degree of the polynomial?

$i$	$x_i$	$f_{i,0}$	$f_{i,1}$	$f_{i,2}$	$f_{i,3}$	$f_{i,4}$	$f_{i,5}$
0	0	132'651	—	—	—	—	—
1	0'2	148'877	81'13	—	—	—	—
2	0'3	157'464	85'87	15'8	—	—	—
3	0'4	166'375	89'11	16'2	1	—	—
4	0'7	195'12	95'79	16'7	1	0	—
5	0'9	216	104'44	17'3	1	0	0

DEGREE 3 OR ↑

b) obtain it using as few data as possible, and evaluate it optimally at  $x=0.25$   
 HORNER

$$P_3(x) = 132.651 + 81.13x + 15.8x(x-0.2) + 1 \cdot x(x-0.2)(x-0.3) =$$

$$= 132.651 + x \left\{ 81.13 + (x-0.2) \left[ 15.8 + (x-0.3) \right] \right\} \quad (\text{now we substitute } x=0.25)$$

$$\underbrace{\hspace{10em}}_{15.75}$$

$$\underbrace{\hspace{8em}}_{0.7875}$$

$$\underbrace{\hspace{6em}}_{81.9175}$$

$$\underbrace{\hspace{4em}}_{20.479375}$$

$$\underbrace{\hspace{2em}}_{153.130375}$$

$$P_3(0.25) = 153.13$$

EXERCISE 13

$x_i$	1.1	1.3	1.4	1.6	1.7
$f(x_i)$	1.669	1.971	2.151	2.577	2.828

Approx  $f(1.35)$  by evaluating the interpol. pol of degree  $\leq 3$  optimally. Estimate the error.  
 (4 or more signif. digits and symm. rounding)

\*AUMENTOS la tabla

$$P_3(x) = 1.669 + 1.51(x-1.1) + 0.9667(x-1.1)(x-1.3) + 0.2667(x-1.1)(x-1.3)(x-1.4) =$$

$$= 1.669 + (x-1.1) \left\{ 1.51 + (x-1.3) \left[ 0.9667 + (x-1.4) \cdot 0.2667 \right] \right\} \quad (\text{now } x=1.35)$$

$$\underbrace{\hspace{10em}}_{-0.01334}$$

$$\underbrace{\hspace{8em}}_{0.04767}$$

$$\underbrace{\hspace{6em}}_{1.5577}$$

$$\underbrace{\hspace{4em}}_{0.389417}$$

$$\underbrace{\hspace{2em}}_{2.0584}$$

$$P(1.35) = 2.0584$$

$$e(x) \approx h_4(x) = 0.25(x-1.1)(x-1.3)(x-1.4)(x-1.6) = \pi(x) \cdot 0.25$$

$$\rightarrow e(x) = f(x) - P_3(x) \rightarrow e(1.35) = f(1.35) - P_3(1.35) \approx P_4(1.35) - P_3(1.35)$$

FINITE DIFFERENCES

DEFINITION:

(FORWARD FIN. DIFF) (+h)

$$\Delta^0 f(x) = f(x)$$

$$\Delta^1 f(x) = \Delta f(x) = f(x+h) - f(x)$$

$$\Delta^k f(x) = \Delta(\Delta^{k-1} f(x)) \quad (k=2, 3, 4, \dots)$$

(k=2)  $\Delta^2 f(x) = \Delta(\Delta f(x)) = \Delta(f(x+h) - f(x)) = f(x+h+h) - f(x+h) - [f(x+h) - f(x)] = f(2h+x) - 2f(x+h) + f(x)$

(We do the same that in the TABLE of DD but without dividing)

TABLE:

	$\Delta^0 f(x_i)$	$\Delta^1 f(x_i)$	$\Delta^2 f(x_i)$	...
$x_0 = x$	$f(x)$	—	—	
$x_1 = x+h$	$f(x+h)$	$f(x+h) - f(x)$	—	
$x_2 = x+2h$	$f(x+2h)$	$f(x+2h) - f(x+h)$	$f(x+2h) - 2f(x+h) + f(x)$	
$\vdots$	$\vdots$	$\vdots$	$\vdots$	

$$\Delta^n f(x) = \Delta^{n-1}(\Delta f(x)) = \Delta^{n-2}(\Delta(\Delta f(x))) = \Delta^{n-3}(\Delta(\Delta(\Delta f(x))))$$

$$\Delta^2(\Delta f(x)) = \Delta^2(f(x+h) - f(x)) = \Delta^3 f(x) \rightarrow \text{which means } \Delta^{n-1}(\Delta f(x)) = \Delta^n f(x)$$

RELATIONSHIP BETWEEN: FINITE DIFF ↔ DIV. DIFF ↔ DERIVS

\*  $f[x_0, x_1, \dots, x_n] = f[x_0, x_0+h, x_0+2h, \dots, x_0+nh] = \frac{\Delta^n f(x_0)}{n! h^n} = \frac{f^{(n)}(\xi)}{n!}$

(OR SOME) BETWEEN THE ROOTS (ξ) btwn. n

↳  $\Delta^n f(x_0) = h^n \cdot f^{(n)}(\xi)$

DIV. DIFF HAS TO BE DIVIDED MANY TIMES

BINOMIAL COEFFICIENTS (NEWTON COEFF)

DEFINITION:

$$\binom{m}{i} = \frac{m!}{i!(m-i)!}$$

↳ m CHOOSE i

Ex  $\binom{5}{3} = \frac{5!}{3!2!} = \frac{5 \cdot 4 \cdot 3 \cdot 2 \cdot 1}{3 \cdot 2 \cdot 1 \cdot 2 \cdot 1} = 10 \rightarrow 10 \text{ ways to divide in subsets of 3}$

- 1 2 3
- 1 2 4
- 1 2 5
- 1 3 4
- 1 3 5
- 1 4 5
- 2 3 4
- 2 3 5
- 2 4 5
- 3 4 5

10

\*  $\binom{t}{n} = \frac{t!}{n!(t-n)!} = \frac{t(t-1)(t-2)\dots(t-n+1)}{n!}$

✓ t ∈ ℝ, n ∈ ℕ

↳ EVEN THOUGH t IS NOT AN INTEGER

$\binom{0}{0} = \frac{0!}{0!0!} = 1 \rightarrow$  PASCAL'S TRIANGLE

$\binom{0}{0} = 1$

$\binom{1}{0} = 1 \quad \binom{1}{1} = 1$

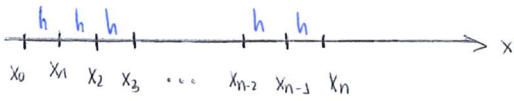
$\binom{2}{0} = 1 \quad \binom{2}{1} = 2 \quad \binom{2}{2} = 1$

$\binom{3}{0} = 1 \quad \binom{3}{1} = 3 \quad \binom{3}{2} = 3 \quad \binom{3}{3} = 1$

EQUALLY  
SPACED

NEWTON POLYNOMIAL WITH EVENLY-SPACED NODES

$$P_n(x) = a_0 + a_1(x-x_0) + a_2(x-x_0)(x-x_1) + \dots + a_{n-1}(x-x_0)(x-x_1)\dots(x-x_{n-2}) + a_n(x-x_0)(x-x_1)\dots(x-x_{n-2})(x-x_{n-1})$$



$$\{x_i = x_0 + ih \quad (i = 0, 1, \dots, n)\}$$

$$x = x_0 + h \cdot t$$

$$t = \frac{x - x_0}{h}$$

$$P_n(x(t)) = q(t) = f[x_0] + f[x_0, x_1] h \cdot t + f[x_0, x_1, x_2] h \cdot h \cdot (t-1) + f[x_0, x_1, x_2, x_3] h \cdot h \cdot (t-1)h(t-2) + \dots$$

(DIV. DIFF.)

$$\dots + f[x_0, x_1, \dots, x_{n-1}] h \cdot h \cdot (t-1) \dots h(t-(n-2)) + f[x_0, x_1, \dots, x_n] h \cdot h \cdot (t-1) \dots h(t-(n-2)) \cdot h(t-(n-1))$$

$$= \Delta^0 f(x_0) + \frac{\Delta^1 f(x_0)}{1! h^1} \cdot h \cdot t + \frac{\Delta^2 f(x_0)}{2! h^2} \cdot h \cdot h \cdot (t-1) + \dots + \frac{\Delta^n f(x_0)}{n! h^n} \cdot h \cdot h \cdot (t-1) \dots h(t-(n-1))$$

(FINITE DIFF.)

$$= \Delta^0 f(x_0) \binom{t}{0} + \Delta^1 f(x_0) \binom{t}{1} + \Delta^2 f(x_0) \binom{t}{2} + \dots + \Delta^n f(x_0) \binom{t}{n} = \sum_{i=0}^n \Delta^i f(x_0) \binom{t}{i} = q(t) = P_n(x(t))$$

(WITH BINOMIAL COEFF.)

HORNER-LIKE (NESTED-PRODUCT) EVALUATION OF q(t)

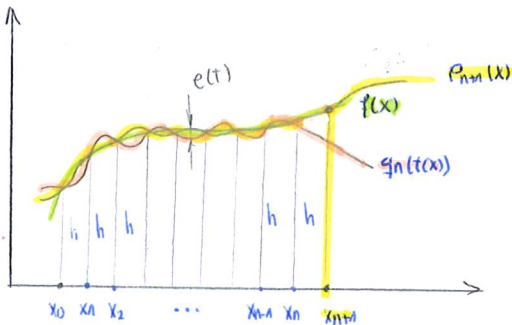
$$q(t) = \Delta^0 f(x_0) \cdot 1 + \Delta^1 f(x_0) \cdot \frac{t}{1} + \Delta^2 f(x_0) \frac{t \cdot (t-1)}{2 \cdot 1} + \dots + \Delta^n f(x_0) \frac{t(t-1)(t-2)\dots(t-(n-2))(t-(n-1))}{n(n-1)(n-2)\dots 3 \cdot 2 \cdot 1}$$

FROM THE FINITE DIFF. TABLE

$$q(t) = \Delta^0 f(x_0) + \frac{t}{1} \left[ \Delta^1 f(x_0) + \frac{(t-1)}{2} \left[ \Delta^2 f(x_0) + \frac{(t-2)}{3} \left( \Delta^3 f(x_0) + \dots \left( \dots + \frac{(t-(n-2))}{n-1} \left( \Delta^{n-1} f(x_0) + \frac{(t-(n-1))}{n} \Delta^n f(x_0) \right) \right) \right) \right] \right]$$

$\underbrace{\hspace{10em}}_{b_{n-1}}$   
 $\underbrace{\hspace{15em}}_{\dots}$   
 $\underbrace{\hspace{20em}}_{b_n}$   
 $b_0 = q(t) = P_n(x(t))$

ESTIMATION OF TRUNCATION ERROR



$$e(t) = f(x(t)) - P_n(x(t)) = f(x(t)) - q_n(t)$$

$$e(t) \approx \Delta^{n+1} f(x_0) \binom{t}{n+1}$$

Adding an extra NODE

PROPAGATION OF A PERTURBATION  $\delta$  IN A TABLE OF FINITE DIFF. (TFD)

$\Delta^0 f(x)$	$\Delta^1 f(x_0)$	$\Delta^2 f(x_0)$	$\Delta^3 f(x_0)$	$\Delta^4 f(x_0)$	$\Delta^5 f(x_0)$	...
$f_0$	—	—	—	—	—	
$f_1$	x	—	—	—	—	
$f_2$	x	x	—	—	—	
$f_3$	x	x	x	—	—	

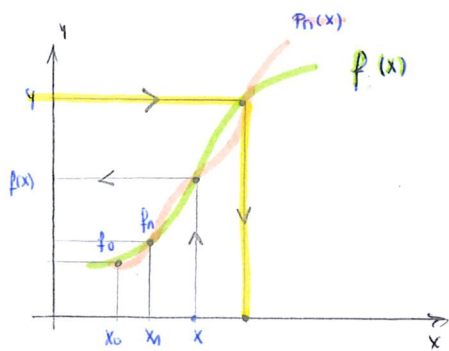
WE WILL ALWAYS HAVE  $\delta$

$f_{4+\delta}$	$x+\delta$	$x+\delta$	$x+\delta$	$x+\delta$	—
$f_5$	$x-\delta$	$x-2\delta$	$x-3\delta$	$x-4\delta$	$x-5\delta$
$f_6$	x	$x+\delta$	$x+3\delta$	$x+6\delta$	$x+10\delta$
$f_7$	x	x	$x-\delta$	$x-4\delta$	$x-10\delta$
$f_8$	x	x	x	$x+\delta$	$x+5\delta$

CONCLUSION

- $\delta$  propagates in a triangle
  - propagates with alternating signs (+/-)
  - abs. values  $\uparrow$  (binomial coeffs)
  - $f \in P_k \rightarrow k$ -th column  $\equiv k!e \rightarrow$  then all  $\delta$
  - $f$  "well behaved", typically get  $\approx 0$  (small values)
- COEFFS. OF THE PASCAL TRIANG.
- \* WELL BEHAVED FUNCTION: smooth,  $\infty$  derivatives

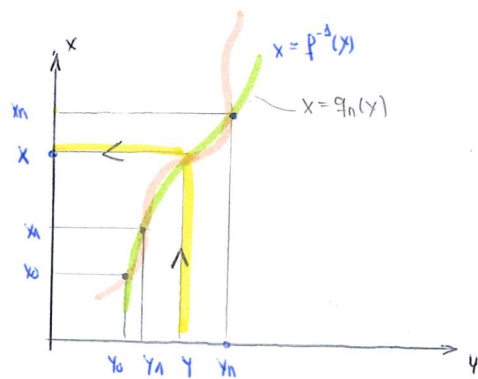
INVERSE INTERPOLATION



(IS GOING TO BE APPROX. BECAUSE WE ARE USING  $P_n(x)$  THAT IS APPROX.)  
 our data now is  $y$   
 $x = f^{-1}(y)$ ?  $\rightarrow$  we want to know the "x" for which we obtain "y"

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

$$P_n(x) = \overset{\text{DATA}}{y} \quad \text{UNKNOWN}$$

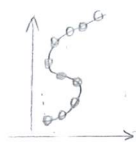


Suppose we have:

$x_i$	$x_0$	$x_1$	$x_2$	...	$x_n$
$y_i$	$y_0$	$y_1$	$y_2$	...	$y_n$

we are changing the axes

\* SI LA FUNCIÓN ES "INEVALUABLE"



$\rightarrow$  YOU CHOSE THE NODES YOU NEED, NOT THE WHOLE ED.

$\rightarrow$  WE TAKE A MONOTONIC PART

ONLY  $\left\{ \begin{array}{l} \text{INCREASING} \\ \text{DECREASING} \end{array} \right.$

EXERCISE 14

$x_i$	0	1	2	3	4	5	6	7	→ EQUALY SPACED → FINITE DIFF
$y_i$	1	2	4	8	15	26	?	?	

a) From a polyn. of what degree?

$x_i$	$\Delta^0 f(x_i)$	$\Delta^1 f(x_i)$	$\Delta^2 f(x_i)$	$\Delta^3 f(x_i)$	$\Delta^4 f(x_i)$
0	1	—	—	—	—
1	2	1	—	—	—
2	4	2	1	—	—
3	8	4	2	1	—
4	15	7	3	1	—
5	26	11	4	1	—
6	(42)	(16)	(5)	(1)	—
7	(64)	(22)	(6)	(1)	—

→ DEGREE 3 OR MORE

b) find "??" without constructing the polynomial

we are going to assume our polynomial is of degree 3

$$\Delta^3 f(x_7) = 1 = \Delta^3 f(x_6)$$

$$\rightarrow \Delta^3 f(x_7) = \Delta^2 f(x_7) - 4 = 1 \rightarrow \Delta^2 f(x_7) = 5 \rightarrow \Delta^3 f(x_6) = \Delta^2 f(x_6) - 5 \rightarrow \Delta^2 f(x_6) = 6$$

$$\rightarrow \Delta^2 f(x_6) = \Delta^1 f(x_6) - 11 \rightarrow \Delta^1 f(x_6) = 16 \rightarrow \Delta^2 f(x_5) = \Delta^1 f(x_5) - 11 \rightarrow \Delta^1 f(x_5) = 22$$

$$\rightarrow \Delta^0 f(x_7) = 26 + 16 = 42 \rightarrow \Delta^0 f(x_8) = 42 + 22 = 64$$

EXERCISE 15

$x_i$	2	4	6	8	10	12	14
$y_i$	?	93	289	569	1071	1813	?

(h=2) (h=2) (h=2) (h=2) (h=2) (h=2)

a) guess "??" (without the polyn.)

$$\left. \begin{array}{l} y_6 = 2843 \\ y_0 = 23 \end{array} \right\} \text{assuming degree} = 3$$

$x_i$	$\Delta^0 f(x_i)$	$\Delta^1 f(x_i)$	$\Delta^2 f(x_i)$	$\Delta^3 f(x_i)$	$\Delta^4 f(x_i)$
2	(23)	—	—	—	—
4	93	(70)	—	—	—
6	259	166	(96)	—	—
8	569	310	144	(48)	—
10	1071	502	192	48	( )
12	1813	742	240	48	0
14	(2843)	(1030)	(288)	(48)	( )

→ DEGREE ≥ 3



EXERCISE 15

b) write the polynomial  $\rightarrow$  CHANGE OF VARIABLE  $x = x_0 + ht$   $\left\{ \begin{array}{l} x = 2 + 2t \\ \hookrightarrow t = \frac{x-2}{2} \end{array} \right.$

$$P_3(x) \rightarrow P_3(x(t)) = q_3(t) = \underbrace{23}_{y_{0,0}} + \underbrace{70}_{y_{1,1}} \binom{t}{1} + \underbrace{96}_{y_{2,2}} \binom{t}{2} + \underbrace{48}_{y_{3,3}} \binom{t}{3}$$

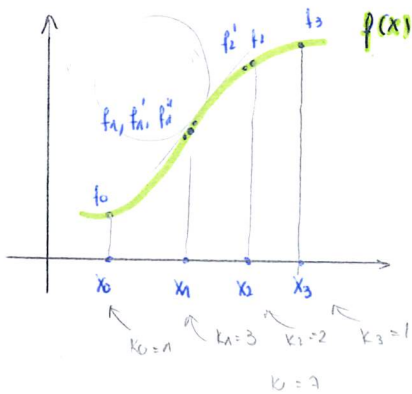
$$q_3(t) = 23 + 70t + 96 \frac{t(t-1)}{2} + 48 \frac{t(t-1)(t-2)}{3 \cdot 2 \cdot 1} = 23 + t \left\{ 70 + \frac{(t-1)}{2} \left[ 96 + \frac{(t-2)}{3} 48 \right] \right\}$$

c) Evaluate it at  $x=5$  optimally

$$x=5 \rightarrow t = \frac{5-2}{2} = 1.5 \rightarrow q_3(1.5) = 23 + 1.5 \left\{ 70 + \frac{0.5}{2} \left[ 96 - \frac{0.5}{3} 48 \right] \right\} = 161$$

$\rightarrow$  que se pega, muy cerca

OSCULATING POLYNOMIALS



$$f^{(m)}(x_i) = f_i^{(m)}$$

we want a polynomial such that:

$$\left\{ \begin{array}{l} P(x_0) = f(x_0); P'(x_0) = f'(x_0); \dots; P^{(m_0)}(x_0) = f^{(m_0)}(x_0) \\ P(x_1) = f(x_1); P'(x_1) = f'(x_1); \dots; P^{(m_1)}(x_1) = f^{(m_1)}(x_1) \\ \dots \\ P(x_n) = f(x_n); P'(x_n) = f'(x_n); \dots; P^{(m_n)}(x_n) = f^{(m_n)}(x_n) \end{array} \right.$$

$m_0, m_1, \dots, m_n$  don't have to be equal

CONDITIONS:

$$P(x_i) = f(x_i); P'(x_i) = f'(x_i); \dots; P^{(m_i)}(x_i) = f^{(m_i)}(x_i) \quad (i=0, 1, \dots, n)$$

$$k_i = (\text{number of conditions at } x_i) = m_i + 1 \rightarrow k_i = m_i + 1$$

$$K = \sum_{i=0}^n k_i \quad \text{NUMBER OF CONDITIONS FOR THE POLYNOMIAL}$$

THEOREM:  $\exists!$   $P_{K-1}(x) \in \mathbb{P}_{K-1}$  that satisfies the osculation interpolation conditions.

TAYLOR POLYNOMIAL



$$f(x) = \underbrace{f(x_0) + f'(x_0)(x-x_0) + \frac{f''(x_0)(x-x_0)^2}{2!} + \dots + \frac{f^{(m)}(x_0)(x-x_0)^m}{m!}}_{\text{EXACT}} + \underbrace{\frac{f^{(m+1)}(\xi)(x-x_0)^{m+1}}{(m+1)!}}_{\text{APPROX}}$$

\* If  $f \in C^{m+1}([x_0, x] \cup [x, x_0])$  then  $\exists \xi \in [x_0, x] \cup [x, x_0]$  /  
and  $f^{(m+1)}$  is defined and bounded in

$$\left\{ \begin{array}{l} \text{EXACT} \rightarrow P_m(x) = f(x_0) + f'(x_0)(x-x_0) + \dots + \frac{f^{(m)}(x_0)(x-x_0)^m}{m!} \\ \text{APPROX} \rightarrow e(x) = \frac{f^{(m+1)}(\xi)(x-x_0)^{m+1}}{(m+1)!} \end{array} \right\} \quad \text{Taylor: } n=0, m_0=m$$

HERMITE POLYNOMIAL (NOT ORTHOGONAL)

\* Any  $n, m_i=1 \quad \forall i (k_i=2i)$

ERROR FOR AN OSCILLATING POLYNOMIAL

$$e(x) = \frac{f^{(k)}(\xi)}{k!} (x-x_0)^{k_0} (x-x_1)^{k_1} \dots (x-x_n)^{k_n}$$

Taylor:  $\begin{cases} k = m+1 = k_0 \\ n=0 \text{ (1 node)} \end{cases}$

Lagrange:  $\begin{cases} n = \text{any} \\ k = n+1 \\ k_i = 1 \end{cases}$

↳ GENERALIZATION OF LAGRANGE TAYLOR

$e(x)$  of a Hermite polynomial  $\rightarrow (k=2(n+1)) = 2n+2$

$$e(x) = \frac{f^{(2n+2)}(\xi)}{(2n+2)!} (x-x_0)^2 (x-x_1)^2 \dots (x-x_n)^2$$

TABLE OF DIVIDED DIFFERENCES WITH REPETITIONS

$x_i$	$f_{i,0}$	$f_{i,1}$	$f_{i,2}$	$f_{i,3}$	$f_{i,4}$	$f_{i,5}$	...
$z_0 = x_0$	$f(x_0) = a_0$	—	—	—	—	—	
$z_1 = x_1$	$f(x_1)$	$\boxed{x} = a_1$	—	—	—	—	
$z_2 = x_1$	$f(x_1)$	$f'(x_1)/1!$	$\boxed{x} = a_2$	—	—	—	
$z_3 = x_2$	$f(x_2)$	$x$	$x$	$\boxed{x} = a_3$	—	—	
$z_4 = x_3$	$f(x_3)$	$x$	$x$	$x$	$\boxed{x} = a_4$	—	
$z_5 = x_3$	$f(x_3)$	$f'(x_3)/1!$	$x$	$x$	$x$	$a_5 = \boxed{x}$	
$z_6 = x_3$	$f(x_3)$	$f'(x_3)/1!$	$f''(x_3)/2!$	$x$	$x$	$x$	$\boxed{\dots}$
$z_7 = x_3$	$f(x_3)$	$f'(x_3)/1!$	$f''(x_3)/2!$	$f'''(x_3)/3!$	$x$	$x$	

$\frac{f^{(n)}(\xi)}{n!}$

so the polynomial will be:

$$P_7(x) = a_0 + a_1(x-z_0) + a_2(x-z_0)(x-z_1) + \dots + a_7(x-z_0)(x-z_1)\dots(x-z_6)$$

### EXERCISE 26

2310212A

a) oscillating pol. /  $f(1)=2$  /  $f'(1)=3$  /  $f(2)=6$  /  $f'(2)=7$  /  $f''(2)=8$

$z_i$	$k_{i0}$	$k_{i1}$	$k_{i2}$	$k_{i3}$	$k_{i4}$
1	2	—	—	—	—
1	2	3	—	—	—
2	6	4	1	—	—
2	6	7	3	2	—
2	6	7	8/2!	1	-1

$$P_4(x) = 2 + 3(x-1) + 1(x-1)^2 + 2(x-1)^2(x-2) - 1(x-1)^2(x-2)^2$$

b) ex)

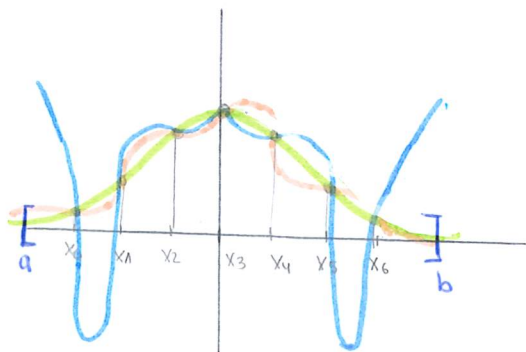
$$e(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \prod(x) \quad \text{f.s. } \xi \text{ between } x \text{ if } f \in C^5$$

$$\hookrightarrow e(x) = \frac{f^{(5)}(\xi)}{5!} \prod(x) = \frac{f[1,1,2,2,2,x]}{5!} \prod(x) = \frac{f^{(5)}(\xi)}{5!} (x-1)^2(x-2)^3$$

### RUNGE EFFECT

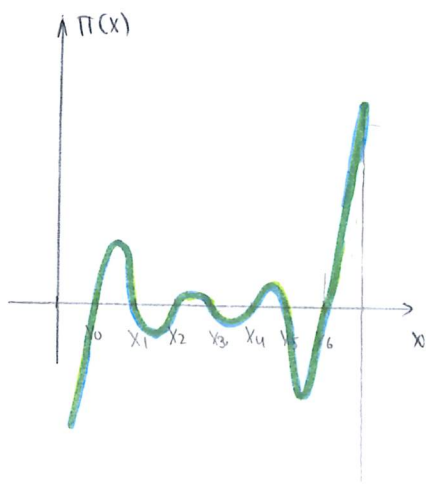
oscillations that come "out of nowhere"

surprising  $\rightarrow$  not in the nature of  $f(x)$



■ The pol. we would expect  
■ Actual pol.

$$e(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} \underbrace{(x-x_0)(x-x_1)\dots(x-x_{n-1})(x-x_n)}_{\prod(x)}$$

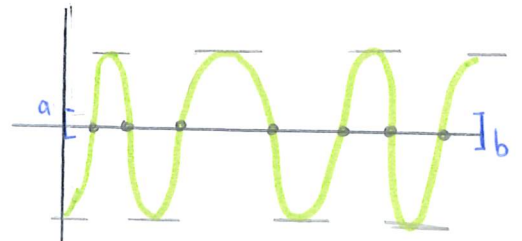


- \* The large oscillations in  $\pi(x)$  translates into errors in our pol. when is equally spaced.
- \* Near the center less oscill.

we will have a max and min and we will try to reach them as many times as possible.

For that, we need to know in which nodes this happens.

↳ Chebyshev  
(WE'LL SEE IT LATER)



### $\|\cdot\|_{\infty}$ DEFINITION (INFINITY NORM)

In  $\mathbb{R}^n$  ;  $\bar{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$

$$\|\bar{x}\|_2 = \sqrt{x_1^2 + x_2^2 + \dots + x_{n-1}^2 + x_n^2}$$

$$\|\bar{x}\|_n = \sqrt[n]{|x_1|^n + |x_2|^n + \dots + |x_n|^n}$$

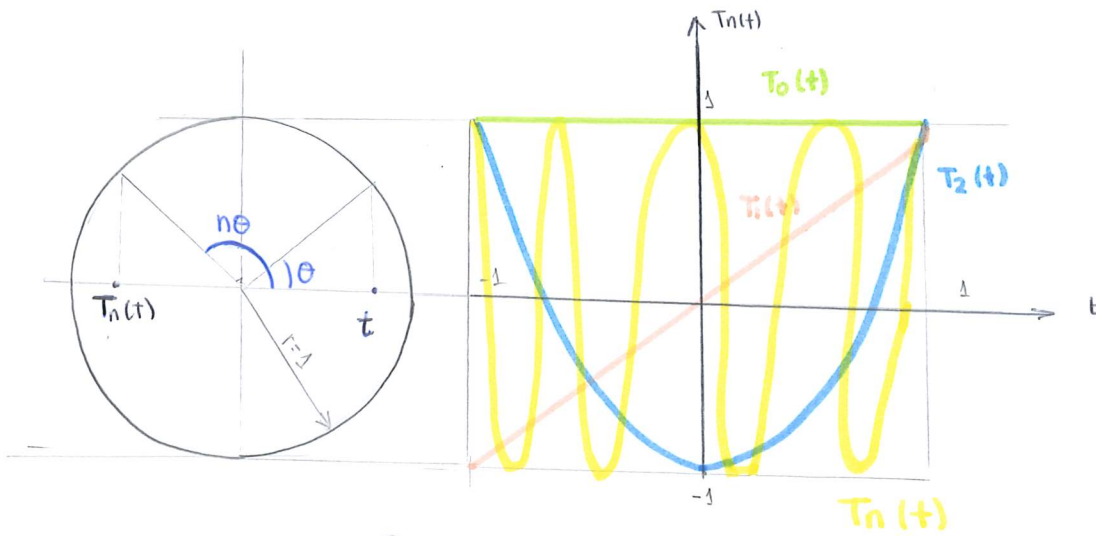
...

$$\|\bar{x}\|_{\infty} = \max_{i=1,2,\dots,n} |x_i| \rightarrow \text{INFINITY NORM} \equiv \text{NORM OF THE MAXIMUM}$$

$$\|f\|_{\infty} = \max_{x \in [a,b]} |f(x)| \rightarrow \text{minimize } \|\pi\|_{\infty}$$

(maximizing the norm minimizes the error)  
↳ the norm of  $\pi(x)$

DEFINITION:  $T_n(t) = \cos(n \underbrace{\arccos(t)}_{\theta}) \quad t \in [-1, 1] \quad (n = 0, 1, \dots)$



$T_0(t) = \cos(0 \cdot \arccos(t)) = 1$

$T_1(t) = \cos(\arccos(t)) = t$

$T_2(t) = \cos(2 \arccos(t)) = 2t^2 - 1$

$T_3(t) = 2t(2t^2 - 1) - t = 4t^3 - 3t$

$T_4(t) = 2t(4t^3 - 3t) - (2t^2 - 1) = 8t^4 - 8t^2 + 1$

→ **RECURSIVE LAW** →  $T_{n+1}(t) = 2t T_n(t) - T_{n-1}(t) \quad (n=1, 2, \dots)$

\* PROOF OF THE RECUR. LAW:  $T_{n+1}(t) = 2t T_n(t) - T_{n-1}(t)$

$\left\{ \begin{aligned} & \cos((n+1)\arccos(t)) = \cos((n+1)\theta) = \cos(\theta + n\theta) = \cos(\theta)\cos(n\theta) - \sin(\theta)\sin(n\theta) \\ & 2\cos\theta \cos(n\theta) - \cos(n\theta - \theta) \dots \end{aligned} \right.$

\* LEADING COEFF =  $2^{n-1} \quad (n=1, 2, \dots)$

ROOTS OF  $T_n(t)$  = "CHEBYSHEV NODES IN  $[-1, 1]$ " IN FACT  $-1$  AND  $1$  WILL NEVER BE A NODE

$T_n(t) = 0 \rightarrow \cos(n\theta) = 0 \rightarrow n\theta = \frac{\pi}{2} + k\pi \quad (k \in \mathbb{Z}) \rightarrow \theta = \frac{\pi/2 + k\pi}{n}$

$$t_i = \cos\theta = \cos\left(\frac{\pi/2 + k\pi}{n}\right) \quad (i=0, 1, 2, \dots, n-1)$$

$$\underbrace{\hspace{10em}}_{n \text{ roots}}$$

$$\cos\left(\frac{(1+2k)\pi}{2n}\right)$$

## LINEAR CHANGE OF VARIABLE $[a,b] \leftrightarrow [-1,1]$

$$x = \frac{a+b}{2} + \frac{(b-a)}{2} t$$

M

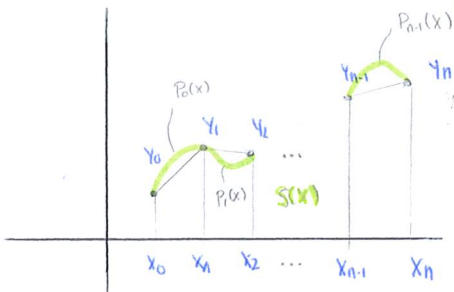
$$J = dx/dt$$

M = midpoint  
J = jacobian

## SPLINES : PIECEWISE POLYNOMIAL FUNCTIONS

24/02/21

\* solution for the Runge effect.



$(x_0 < x_1 < x_2 \dots < x_{n-1} < x_n) \rightarrow$  INTERPOL. POINTS : SORTED AND DIFFERENT

• If using pol. of degree 1 we won't have a continuous 1st derivative.

• our approx  $\rightarrow$  pol of degree 3  $\rightarrow$  MOST OF THE TIMES IN THE EXERCISES (3 or ODD)

$$P_i(x) = a_0 + a_1 x + a_2 x^2 + a_3 x^3$$

$\rightarrow a_0, a_1, a_2, a_3 \Rightarrow 4 \text{ dof}$

\* A spline of order  $l$  is a piecewise pol. function  $s(x)$  that satisfies:

$$s(x_i) = f(x_i) \quad (i=0,1,\dots,n)$$

PIECW. APPROX = ACTUAL POL.

$$s(x)|_{[x_i, x_{i+1}]} = P_i(x) \in \mathbb{P}_l$$

$$s \in C^{l-1}(\mathbb{R}) \rightarrow \text{SMOOTHNESS}$$

\* if  $l=3$   
CUBIC

$$s(x)|_{[x_i, x_{i+1}]} = P_i(x) \in \mathbb{P}_3$$

$$s \in C^2(\mathbb{R}) \text{ (quite smooth)}$$

## NUMBER OF DEGREES OF FREEDOM VS. CONDITION

$$n \text{ dof} = n$$

$$n \text{ conds} : 2n \text{ (interpolate)}$$

$$n-1 \quad (C^1)$$

$$n-1 \quad (C^2)$$

$$4n-2$$

$\rightarrow$  2 dof that we can use freely

CUBIC SPLINE WITH BOUNDARY CONDITIONS

$$\begin{cases} P_0'(x_0) = f'(x_0) \\ P_{n-1}'(x_n) = f'(x_n) \end{cases}$$

\* NOT - a - knot condition: 3rd der. is continuous (we are not doing that)

NATURAL CUBIC SPLINE

$$\begin{cases} P_0''(x_0) = 0 \\ P_{n-1}''(x_n) = 0 \end{cases}$$

PERIODIC CUBIC SPLINE

\* Only if  $f(x)$  is PERIODIC  
NOT for the exercises

$$\begin{cases} P_0'(x_0) = P_{n-1}'(x_n) \\ P_0''(x_0) = P_0''(x_n) \end{cases}$$

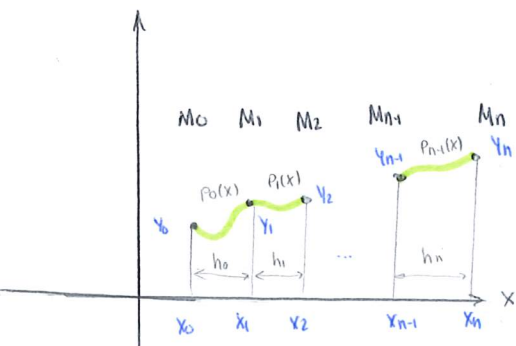
OPTIMAL POL. // THEOREM OF OPTIMALITY

\*  $\forall g \in C^2([x_0, x_n]) / g(x_i) = f(x_i) \ (i=0,1,\dots,n), \int_{x_0}^{x_n} g''(x)^2 dx \geq \int_{x_0}^{x_n} s''(x)^2 dx$   
CUBIC SPLINE WITH BOUNDARY COND  
 $g'(x_0) = f'(x_0)$   
 $g'(x_n) = f'(x_n)$

\*  $\forall g \in C^2([x_0, x_n]) / g'(x_i) = f'(x_i) \ (i=0,1,\dots,n), \int_{x_0}^{x_n} g''(x)^2 dx \geq \int_{x_0}^{x_n} s''(x)^2 dx$   
NATURAL CUBIC SPLINE  
 $g'(x_0) = 0$   
 $g''(x_n) = 0$  } The one with the least amount of oscillations

\*  $\forall g \in C^2([x_0, x_n]) / g'(x_i) = f'(x_i) \ (i=0,1,\dots,n), \int_{x_0}^{x_n} g''(x)^2 dx \geq \int_{x_0}^{x_n} s''(x)^2 dx$   
PERIODIC CUBIC SPLINE  
 $g'(x_0) = g'(x_n)$   
 $g''(x_0) = g''(x_n)$

MOMENTS



\*  $M_i = S''(x_i)$  and find them 4ST  
 $(n-1) \cdot (n-1)$   
 $(n+1) \cdot (n+1)$

\* the matrix will be triangular and strictly diagonally dominant.

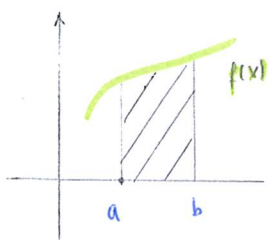
$$\begin{pmatrix} a_{11} & a_{12} & \dots \\ a_{21} & a_{22} & a_{23} \\ \dots & \dots & \dots \end{pmatrix} \quad \begin{matrix} a_{11} > a_{12} \\ a_{22} > a_{21} + a_{23} \\ \dots \end{matrix}$$

$$\begin{pmatrix} 2h_0 & h_0 & \dots & 0 & 0 \\ h_0 & 2(h_1+h_2) & h_1 & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 2(h_{n-2}+h_{n-1}) & h_{n-1} \\ 0 & 0 & \dots & h_{n-1} & 2h_{n-1} \end{pmatrix} \begin{pmatrix} M_0 \\ M_1 \\ \vdots \\ M_{n-1} \\ M_n \end{pmatrix} = \begin{pmatrix} d_0 \\ d_1 \\ \vdots \\ d_{n-1} \\ d_n \end{pmatrix}$$

(T)



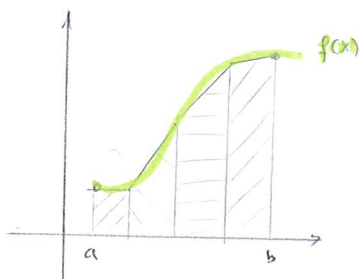
# CH. 2. INTEGRATION



$$\int_a^b f(x) dx = F(b) - F(a) \quad \text{if } F' = f$$

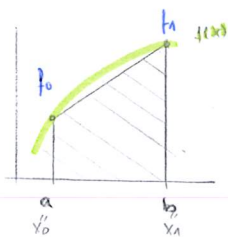
BURROW'S LAW

## COMPOUND TRAPEZOIDAL RULE



addition of the areas (trapeziums)

\* If we don't divide it into intervals:



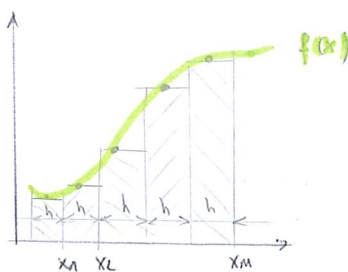
$$\frac{(b-a)(f_0+f_1)}{2} = \underbrace{\left(\frac{h}{2}\right)}_{w_0} f_0 + \underbrace{\left(\frac{h}{2}\right)}_{w_1} f_1$$

WEIGHT FUNCTIONS

$$Q = \frac{h}{2} (f_0 + f_1)$$

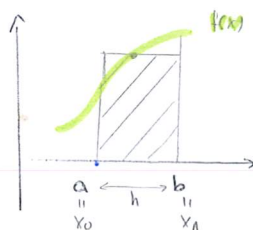
## COMPOUND MIDPOINT RULE

03/03/21



we take the MP. of each interval and obtain areas (square (rectang)) that we will add.

\* If we don't divide it into intervals:



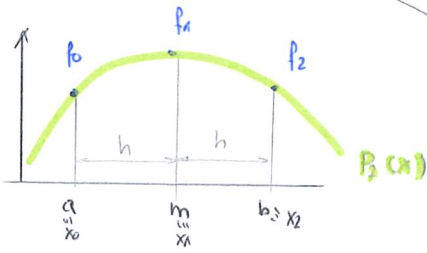
$$Q = h \cdot f_0$$

$$Q \approx \int_a^b f(x) dx \quad (\text{QUADRATURE FUNCT})$$

QUADRATURE RULE

$$I = \int_a^b f(x) dx \approx Q$$

→ Instead of integrating our function, we will integrate our Lagrange interpolation polynomial.



(Q17)

$$I = \int_a^b P_2(x) dx = Q$$

EXAMPLE

(Q55)

SIMPSON'S RULE VIA INTEGRATION OF LAGRANGE BASED FUNCTIONS

$$I = \int_a^b f(x) dx \approx \int_a^b P_2 dx = Q \quad (\text{QUAD. RULE})$$

$$P_2(x) = \underbrace{\frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)}}_{L_0(x)} f_0 + \underbrace{\frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)}}_{L_1(x)} f_1 + \underbrace{\frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)}}_{L_2(x)} f_2 \quad \left. \vphantom{P_2(x)} \right\} \text{LAGR. POL.}$$

$$* Q = \int_a^b P_2 dx = \int_a^b (L_0(x)f_0 + L_1(x)f_1 + L_2(x)f_2) dx = \underbrace{\left( \int_a^b L_0(x) dx \right)}_{W_0} f_0 + \underbrace{\left( \int_a^b L_1(x) dx \right)}_{W_1} f_1 + \underbrace{\left( \int_a^b L_2(x) dx \right)}_{W_1} f_2$$

THEOREM:

$$P_n(x) = \sum_{i=0}^n L_i(x) f_i$$

$$Q = \sum_{i=0}^n W_i f_i \quad \text{where } W_i = \int L_i(x) dx \quad (i=0, 1, \dots, n)$$

CHANGE OF VARIABLE

$$x = x_0 + ht \quad \text{for } w_0 = \int_{x_0}^b L_0(x) dx = \int_{x_0}^{x_2} \frac{(x-x_0)(x-x_2)}{h} dx$$

\* Applying the change of variable:

$$x = x_1 + ht \rightarrow x - x_2 = \frac{x_1 - x_2 + ht}{-h}$$

$$w_1 = \int_{-1}^1 \frac{K(t+1)K(t-1)}{K(-1)} \cdot h dt = \frac{4h}{3} \rightarrow \boxed{w_1 = \frac{4h}{3}}$$

$w_0 = h/3$  ——— TYPICAL VALUES WE ARE GOING TO USE

Q<sub>SS</sub> VIA NEWTON POLYNOMIAL (same as before but using Newton interp. pol.)

$$Q_{SS} = \int_{x_0}^{x_2} p_2(x) dx = \int_0^2 \underbrace{p_2(x(t))}_{q_2(t)} h dt$$

$$+ q_2(t) = f(x_0) + \Delta f(x_0) \binom{t}{1} + \Delta^2 f(x_0) \binom{t}{2} = f_0 + (f_1 + f_0)t + (f_2 - 2f_1 + f_0) \frac{t(t-1)}{2} \quad \left. \vphantom{q_2(t)} \right\} \text{NEWTON POL}$$

$$Q_{SS} = \int_0^2 \left[ f_0 + (f_1 - f_0)t + (f_2 - 2f_1 + f_0) \frac{t^2 - t}{2} \right] h dt = h \left( 2f_1 + \frac{f_2 - 2f_1 + f_0}{3} \right) = \left( \frac{f_0}{3} + \frac{4}{3}f_1 + \frac{f_2}{3} \right) h$$

$$\rightarrow \boxed{Q_{SS} = \frac{h}{3} (f_0 + 4f_1 + f_2)}$$

$$\rightarrow f_0 \left( \frac{h}{3} \right) + f_1 \left( \frac{4h}{3} \right) + \dots$$

$\underbrace{\hspace{10em}}_{w_0 \quad w_1}$   
SAME AS BEFORE

POLYNOMIAL DEGREE OF A QUADRATURE RULE

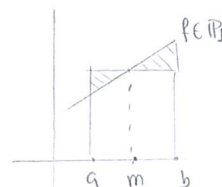
Polynomial degree (of exactitude) or "order" of a quadrature rule

$$\boxed{N \geq n = n_n - 1}$$

\* In the compound trapezoidal rule: we integrate polynom. of degree 0 and 1 so:  $\boxed{N=1}$

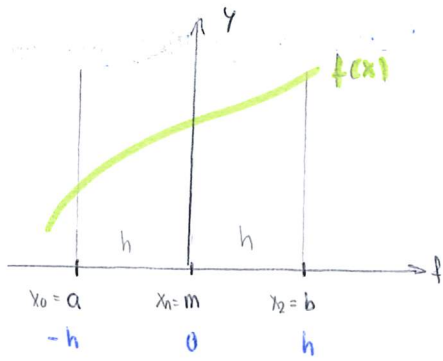
\* In the compound MIDPOINT rule: we integr. pol. of degree 0 and luckily also of degree 1:  $\boxed{N=1}$

\* In the SIMPSON'S RULE  $\rightarrow \boxed{N=3}$



INDETERMINATE COEFFICIENTS. (QPS VIA INDET. COEF).

- 1) AXE "X" + FUNCTION + NODES
- 2) PLACE AXE "Y"
- 3) IMPOSE THE EXACT INTEG.



\* EXACT INTEGRATION OF:

$$f(x) \equiv 1 \rightarrow \int_{-h}^h 1 dx = 2h = \overset{N \geq 2}{w_0 \cdot 1 + w_1 \cdot 1 + w_2 \cdot 1}$$

$$f(x) \equiv x \rightarrow \int_{-h}^h x dx = 0 = w_0(-h) + w_1(0) + w_2(h) = \underline{(w_2 - w_0)h = 0}$$

$$f(x) \equiv x^2 \rightarrow \int_{-h}^h x^2 dx = \frac{2h^3}{3} = w_0 \cdot h^2 + w_1(0) + w_2 h^2 = \underline{(w_0 + w_2)h^2 = \frac{2h^3}{3}}$$

So we have:

$$\begin{cases} \bullet w_0 + w_1 + w_2 = 2h \\ \bullet w_2 = w_0 \\ \bullet w_0 + w_2 = 2h/3 \end{cases} \rightarrow \begin{cases} w_1 = 2h - 2h/3 = 4h/3 \\ w_0 = w_2 = h/3 \end{cases} \text{ WHAT WE EXPECTED}$$

TRUNCATION ERROR (FOR INT.)

$$\underbrace{\int_a^b f(x) dx}_I = \underbrace{\int_a^b p_n(x) dx}_Q + E$$

$$\rightarrow E = I - Q$$

EXACT (I)      APPROX (Q)

→ The error in the integral is the same as the int. of the error.

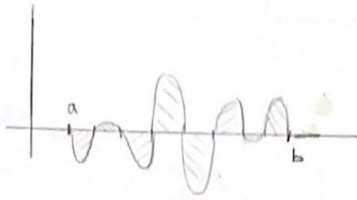
$$* E = \int_a^b f(x) dx - \int_a^b p_n(x) dx = \int_a^b (f(x) - p_n(x)) dx = \int_a^b e(x) dx$$

$$\int_a^b e(x) dx = \int_a^b f[x_0, x_1, \dots, x_{n+1}] \pi(x) dx = \int_a^b \frac{f^{(n+1)}(\xi(x))}{(n+1)!} \pi(x) dx$$

\* case:  $f \in \mathbb{P}_m$  with central symm. of nodes AND ODD NUMBER OF NODES

$f^{(N+1)}(\xi) = k \cdot \xi$

$$E = f^{(N+1)}(\xi) \int_a^b \pi(x) dx = 0$$



all of them cancel out  $\rightarrow \int_a^b \pi(x) dx = 0$

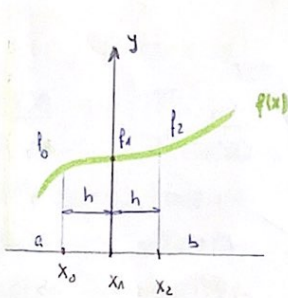
$f \in \mathbb{P}_n, E=0 (N \geq n)$

$f \in \mathbb{P}_{nm}, E=0 (N \geq nm)$  (odd + central symm)

POI DEGREE  $(N+1)$   
 $E = k \cdot f^{(N+1)}(\xi) \quad f.s. \xi \in [a, b]$  GENERAL EXPRESSION

- $f \in \mathbb{P}_N \rightarrow f^{(N+1)}(\xi) = 0 \rightarrow E = 0$
- $f \in \mathbb{P}_N \quad \text{deg}(f) = N+1 \rightarrow f^{(N+1)}(\xi) = k \cdot \xi \rightarrow E \neq 0$

TRUNCATION ERROR OF SIMPLE SIMPSON'S RULE ( $E_{SS}$ )



INDET. COEFF

$$\begin{cases} \int_{-h}^h 1 dx = Q \\ \int_{-h}^h x dx = Q \\ \int_{-h}^h x^2 dx = Q \end{cases} \rightarrow Q = \frac{h}{3} (f_0 + 4f_1 + f_2) \quad \begin{matrix} N \geq 2 \\ N = 3 \end{matrix}$$

$\int_{-h}^h x^3 dx = Q + E^0 = Q + I \rightarrow Q = \frac{h}{3} ((x^3) + 4 \cdot 0 + h^3) = E \rightarrow E = 0$

$f(x) = x^{N+1} = x^4 \rightarrow \int_h^h x^4 dx = Q + E \rightarrow \frac{2h^5}{5} = \frac{h}{3} \left( \frac{(h)^4 + 4 \cdot 0 + h^4}{2h^4} \right) + \frac{k f^{(4)}(\xi)}{E} \quad f.s. \xi \in [a, b]$

$f = x^4 \Rightarrow f^{(4)} = 4! = 4 \cdot 3 \cdot 2 = 24$   
 $f' = 4x^3$   
 $f'' = 12x^2$   
 $f''' = 24x$   
 $f^{(4)} = 24$  KFE

$$k = \frac{2h^5}{5} - \frac{2h^5}{3} + (24)k = \frac{h^5}{12} \left( \frac{1}{5} - \frac{1}{3} \right) = \frac{-h^5}{6 \cdot 15} = \frac{-h^5}{90}$$

$$E_{55} = \frac{-f^{(4)}(\xi)}{90} \cdot h^5$$

INTERPOLATORY QUADRATURE RULE CLASSIFICATION

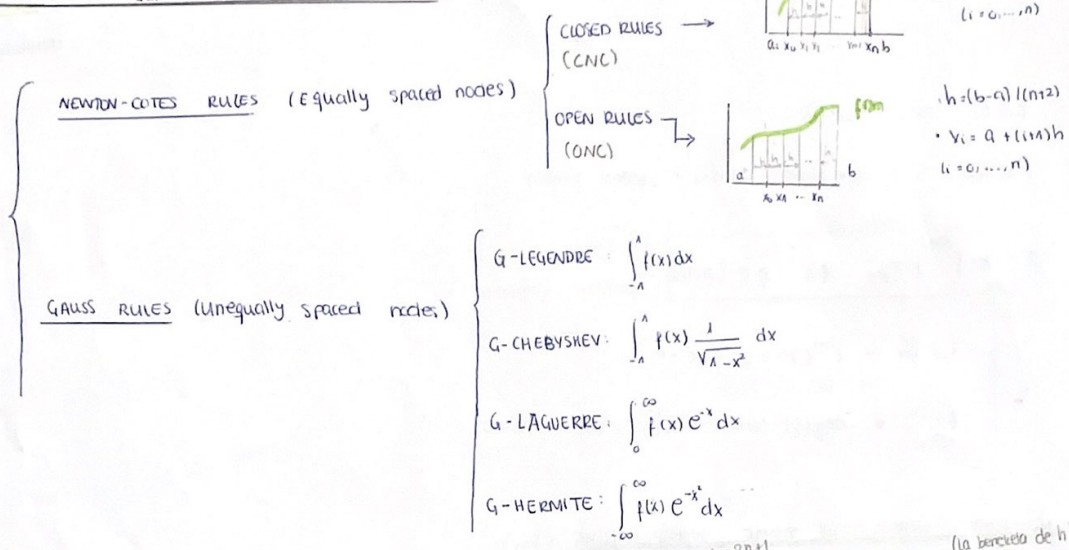


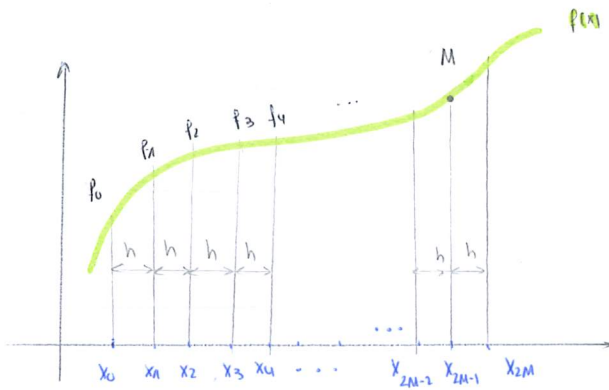
TABLE OF NEWTON-COTES (NC)

NAME	CLASSIF	NODES nn	Q	$(nn-1) = n$	$N \geq n$	$E_s$	$N+2 =$	OCS	$E_c$	OCC
Mid Point	ONC	1	$2h f_0$	0	1	$h^3 f''(\xi)/3$	3		$h^2(b-a) f''(\xi)/6$	2
Trapezoidal	CNC	2	$\frac{h}{2} (f_0 + f_n)$	1	1	$-h^3 f'''(\xi)/12$	3		$-h^2(b-a) f'''(\xi)/12$	2
Simpson	CNC	3	$\frac{h}{3} (f_0 + 4f_1 + f_2)$	2	3	$-f^{(4)}(\xi) h^5/90$	5		$-h^4(b-a) f^{(4)}(\xi)/180$	4
2nd Simpson	CNC	4	$\frac{3h}{8} (f_0 + 3f_1 + 3f_2 + f_3)$	3	3	$-3h^5 f^{(5)}(\xi)/180$	5		$-h^4(b-a) f^{(4)}(\xi)/180$	4
Milne/Boole	CNC	5	$\frac{2h}{45} (7f_0 + 32f_1 + 12f_2 + 32f_3 + 7f_4)$	4	5	$-8h^7 f^{(7)}(\xi)/945$	7		$-2h^4(b-a) f^{(4)}(\xi)/945$	6
	ONC	2	$\frac{3h}{2} (f_0 + f_n)$	1	1	$3h^3 f'''(\xi)/4$	3		$h^2(b-a) f'''(\xi)/4$	2
	ONC	3	$\frac{4h}{3} (2f_0 - f_1 + 2f_2)$	2	3	$14h^5 f^{(5)}(\xi)/45$	5		$7h^4(b-a) f^{(4)}(\xi)/90$	4
	CNC	9	$\frac{h}{14475} (3956f_0 + \dots - 2f_8 + \dots)$	8	9	...	10		...	9

Notes:  $N = 2n+1$ ,  $N = nn$  (impair),  $f, f' \in C[a, b]$ ,  $N+2 =$  (la benche de h),  $E_c$  (la benche de h),  $E_s$  (la benche de h),  $OCS$  (la benche de h),  $OCC$  (la benche de h).

Es COEFF: ONC (+)  
 CNC (-)

Big O = OCS  $\rightarrow$  ex  $\frac{14h^5 f^{(5)}(\xi)}{45} = O(h^5)$



• M equally-wide subintervals

$$(b-a) = M \cdot h \rightarrow h = \frac{b-a}{2M}$$

$$Q_{cs} = \frac{h}{3} (f_0 + 4f_1 + f_2) + \frac{h}{3} (f_2 + 4f_3 + f_4) + \frac{h}{3} (f_4 + 4f_5 + f_6) + \dots$$

CODING

function Q = compound\_simpson(f, a, b, ns)

$$h = (b-a) / ns / 2;$$

$$xi = linspace(a, b, 2 * ns + 1);$$

$$Q = h/3 * (f(a) + f(b)) + \dots$$

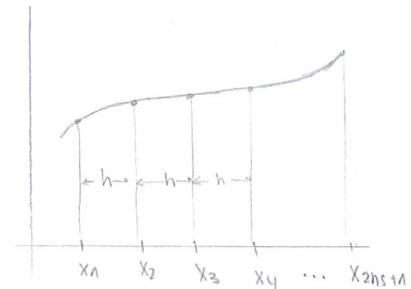
$$+ 4 * h/3 * \text{sum}(f(xi(2:2:end-1))) + \dots$$

INTERIOR EVEN NODES (IN ODD NS)

$$+ 2 * h/3 * \text{sum}(f(xi(3:2:end-2)))$$

function handed

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TRUNCATION ERROR OF THE COMPOUND SIMPSON'S RULE (Ecs)

(Algebraic sum)

$$E_{cs} = \sum_{i=1}^M \left( -\frac{h^5}{90} f^{(4)}(\xi_i) \right) = E_{ss,i} \quad (f.s. \xi_i \in \text{ith-subinterval if } f \in C^4([a,b]))$$

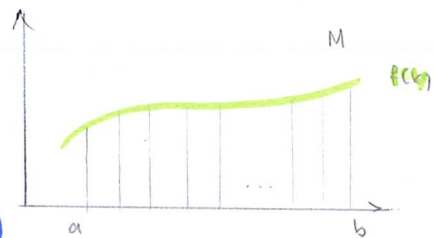
$$= -\frac{h^5}{90} \frac{M \sum_{i=1}^M f^{(4)}(\xi_i)}{M} = \left( \exists \xi \in [a,b] / f^{(4)}(\xi) = \overline{f^{(4)}} \right)$$

$$= -\frac{h^5}{90} \cdot \frac{(b-a)}{2h} \cdot \overline{f^{(4)}} = \leftarrow \text{WEIERSTRASS'S INTERMEDIATE VALUE}$$

$$= -\frac{h^4}{180} (b-a) f^{(4)}(\xi)$$

$O(h^4)$

$$|E_{cs}| = \left| -\frac{h^4}{180} (b-a) f^{(4)}(\xi) \right| \dots$$



ERROR PROPAGATION. AMPLIFICATION FACTOR (AF)

$Q = \sum_{i=0}^n w_i f(x_i) = \sum_{i=0}^n w_i f_i$

\* Instead of  $f_i$  we have perturbed values  $\bar{f}_i$  ("all else equal / exact")

$\bar{Q} = \sum_{i=0}^n w_i \bar{f}_i \quad |f_i - \bar{f}_i| \leq \epsilon \forall i$

$|E_r| = |Q - \bar{Q}| = \left| \sum_{i=0}^n w_i f_i - \sum_{i=0}^n w_i \bar{f}_i \right| = \left| \sum_{i=0}^n w_i (f_i - \bar{f}_i) \right| \leq \sum_{i=0}^n |w_i| |f_i - \bar{f}_i| \leq |E_r| \leq \epsilon \left( \sum_{i=0}^n |w_i| \right)$  AF

$\int_a^b f(x) dx = \int_a^b 1 dx = b-a = Q = \sum_{i=0}^n w_i f(x_i) = \sum_{i=0}^n w_i \cdot 1 = \sum_{i=0}^n w_i = \underline{\underline{b-a}}$

$(b-a) = \sum_{i=0}^n w_i = \sum_{i=0}^{n^+} w_i + \sum_{i=0}^n -w_i = \sum^+ |w_i| - \sum^- |w_i|$

$AF = \sum |w_i| = \sum^+ |w_i| + \sum^- |w_i| = (b-a) + \sum^- |w_i| + \sum^+ |w_i| = \boxed{(b-a) + 2 \sum^- |w_i|} = AF$

\* If all  $w_i > 0 \rightarrow AF = b-a$  and  $Q$  is said to be "STABLE".

THEOREM ABOUT CONVERGENCE OF QUADR. RULES WITH POSITIVE COEFFICIENTS

\* Assume there is an algorithm generating sets of  $1, 2, 3, \dots$  nodes.

- $\{x_0^0\}$
- $\{x_0^1, x_n^1\}$
- $\{x_0^2, x_n^2, x_2^2\}$
- ...
- $\{x_0^n, x_n^n, x_2^n, \dots, x_n^n\}$
- ...

•  $f \in C([a, b])$

• And also assume all corresponding interpolatory weights are  $> 0 \rightarrow w_i > 0$

• Call  $Q^n = \sum_{i=0}^n w_i^n f(x_i^n)$

$\hookrightarrow$  Then:  $\lim_{n \rightarrow \infty} Q^n = I = \int_a^b f(x) dx$  (converges to  $I$ )

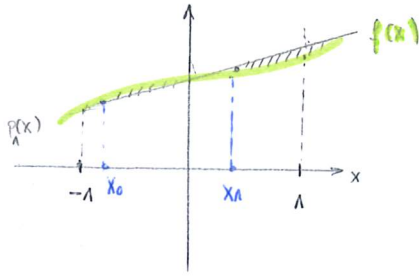
SIMILAR THEOREM WITH COMPOUND RULES

If  $f \in C([a, b])$ , calling  $Q_C^M$  the value given by a compound rule with positive coefficients and  $M$  subintervals, then:

$\lim_{M \rightarrow \infty} Q_C^M = I = \int_a^b f(x) dx$



GAUSS-LEGENDRE RULE OF 2 NODES (GAUSS QUADRATURE)



• nodes: not necessarily equally spaced.

$$\int_{-1}^1 f(x) dx ;$$

$$* Q_{GL2} = Q = \sum_{i=0}^1 w_i f(x_i) = w_0 f(x_0) + w_1 f(x_1)$$

$$\begin{cases} n=1 \\ N \geq 3 \end{cases} ; \boxed{N = 3 = 2n+1} = \frac{2n+1}{2} \text{ nodes}$$

(4 degrees of freedom  $\begin{cases} w_0, w_1 \\ x_0, x_1 \end{cases}$ )

(EXACT INT)

EXACT INT. OF  $f(x) = 1$  :  $\int_{-1}^1 1 dx = 2 = w_0 \cdot 1 + w_1 \cdot 1 + \dots$

EXACT INT. OF  $f(x) = x$  :  $\int_{-1}^1 x dx = 0 = w_0 \cdot x_0 + w_1 \cdot x_1$

EXACT INT. OF  $f(x) = x^2$  :  $\int_{-1}^1 x^2 dx = \frac{2}{3} = w_0 x_0^2 + w_1 x_1^2$

EXACT INT. OF  $f(x) = x^3$  :  $\int_{-1}^1 x^3 dx = 0 = w_0 x_0^3 + w_1 x_1^3$

SOL:  $w_0 = w_1 = 1 / x_0 = -1/\sqrt{3} / x_1 = 1/\sqrt{3}$   
 $w_0 = w_1 = 1 / x_0 = 1/\sqrt{3} / x_1 = -1/\sqrt{3}$  } BOTH PRODUCE THE EXACT SAME QUADR. RULE

$$Q = 1 \cdot f(-1/\sqrt{3}) + 1 \cdot f(1/\sqrt{3})$$

→ FORMALLY : 2 diff. sol  
 → 1 only  $\phi$

GAUSS-LEGENDRE RULE OF 3 NODES

(6 degrees of freedom  $\begin{cases} w_0, w_1, w_2 \\ x_0, x_1, x_2 \end{cases}$ )

EXACT INT OF :

$f(x) = 1$  :  $\int_{-1}^1 1 dx = 2 = w_0 \cdot 1 + w_1 \cdot 1 + w_2 \cdot 1$

$f(x) = x$  :  $\int_{-1}^1 x dx = 0 = w_0 x_0 + w_1 x_1 + w_2 x_2$

$f(x) = x^2$  :  $\int_{-1}^1 x^2 dx = \frac{2}{3} = w_0 x_0^2 + w_1 x_1^2 + w_2 x_2^2$

$f(x) = x^3$  :  $\int_{-1}^1 x^3 dx = 0 = w_0 x_0^3 + w_1 x_1^3 + w_2 x_2^3$

$f(x) = x^4$  :  $\int_{-1}^1 x^4 dx = \frac{2}{5} = w_0 x_0^4 + w_1 x_1^4 + w_2 x_2^4$

$f(x) = x^5$  :  $\int_{-1}^1 x^5 dx = 0 = w_0 x_0^5 + w_1 x_1^5 + w_2 x_2^5$

SOL: WE WOULD OBTAIN 6 DIFF. SOLUTIONS, FORMALLY → SAME Q FOR ALL.

$$\begin{cases} n=2 \\ N=5 = \frac{2n+1}{2} \text{ nodes} \end{cases}$$

\*  $n+1$  nodes  $\{ x_0, \dots, x_n \}$  ;  $n$  dif =  $2(n+1) = 2n+2$  CONDITIONS and  $N = 2n+1$

• REAL, DISTINCT AND IN THE INTERVAL  $(-b, b)$

# SCALAR PRODUCT AND ORTHOGONAL POLYNOMIALS OF FUNCTIONS

• SCALAR PRODUCT:  $\langle f, g \rangle = \int_a^b f(x)g(x)w(x)dx$   $w(x) \geq 0 \quad \forall x \in (a,b)$

• ORTHOGONALITY:  $f, g$  orthogonal  $(f \perp g) \iff \langle f, g \rangle = 0$

• Legendre pols  $\{ \varphi_0(x), \varphi_1(x), \dots, \varphi_n(x), \dots \}$  are orthogonal in  $[-1, 1]$  wr to  $w(x) \equiv 1$

$$\int_{-1}^1 \varphi_i(x)\varphi_j(x)dx = 0 \quad \forall i, j / i \neq j$$

DONT NEED TO KNOW HOW TO APPLY IT ON EXERCISES

\* we are going to orthogonalize using Gram Schmidt method (ALGEBRA)

$\{ 1, x, x^2, x^3, \dots \}$  } ORTHOGONALIZE

$\{ 1, \varphi_1(x), \varphi_2(x), \dots \}$  } NORMALIZE  $\rightarrow$  most usual normalization  $\varphi_i(1) = 1$

$\{ 1, x, \frac{3}{2}x^2 - \frac{1}{2}, \frac{5}{2}x^3 - \frac{3}{2}x, \dots \}$

THE MOST IMPORTANT THING IS THAT THEY ARE ORTHOGONAL

• ROOTS:  $x = \pm\sqrt{\frac{1}{3}}$ ;  $x = \pm\sqrt{\frac{3}{5}}$  ...  $\rightarrow$  same as GAUSS-LEGENDRE RULE!!

Ex)  $\{ 1, x, \frac{3}{2}x^2 - \frac{1}{2} \} \equiv$  Base of  $\mathbb{P}_2$  ORTHOGONAL to all polynomials of degree equal or less than itself.

— ORTHOGONAL POL.  
— FAMILY OF ORTHOG.

## FUNDAMENTAL THEOREM OF GAUSS-LEGENDRE QUADRATURE

If  $f \in \mathbb{P}_{2n+1}$  and  $Q_{GL}$  is the <sup>interpolatory</sup> quadrature rule with nodes equal to the Legendre polynomial  $\varphi_{2n+1}(x)$  of degree  $2n+1$ , then  $f$  is integrated exactly by  $Q_{GL}$ :  
(that rule)

$$\int_{-1}^1 f(x)dx = \sum_{i=0}^n w_i f(x_i) + \text{X}$$

PROOF:

Let  $f$  be any polynomial of degree  $\leq 2n+1$ .

Divide it by  $\varphi_{2n+1}(x)$  (Leg. pol of degree  $2n+1$ )

$$P_{2n+1}(x) = \frac{\Psi_{2n+1}(x)}{\Psi_n(x)} = \frac{a_n x^n + a_{n-1} x^{n-1} + \dots + a_2 x^2 + a_1 x + a_0}{\Psi_n(x)} \in \mathbb{P}_n$$

↑ EVERY TIME ADDING A MONOMIAL,  
WE REDUCE THE DEGREE OF THE REMAINDER

$$P_{2n+1}(x) = q_n(x) \cdot \Psi_n(x) + \Psi_n(x)$$

IT CAN BE 0

$$\int_{-1}^1 P_{2n+1}(x) dx = \int_{-1}^1 q_n(x) \Psi_n(x) dx + \int_{-1}^1 \Psi_n(x) dx = \int_{-1}^1 \Psi_n(x) dx = \sum_{i=0}^n W_i \cdot \Psi(x_i)$$

FOR EVERY INTERP. Q OF  $n+1$  NODES  
FOR ANY

Choose the nodes:  $x_i = \xi_i =$  roots of  $\Psi_n(x)$

$$P_{2n+1}(\xi_i) = q_n(\xi_i) \underbrace{\Psi_n(\xi_i)}_{\text{IS A ROOT}} + \Psi_n(\xi_i) \rightarrow P_{2n+1}(\xi_i) = \Psi_n(\xi_i)$$

$$\int_{-1}^1 P_{2n+1}(x) dx = \sum_{i=0}^n W_i P_{2n+1}(\xi_i)$$

Q412 VIA LEGENDRE POLYNOMIAL  $\Psi_2(x)$

$$\Psi_2(x) = \frac{3x^2}{2} - \frac{1}{2}$$

1ST) ROOTS OF  $\Psi_2(x) =$  NODES

$$\frac{3x^2}{2} - \frac{1}{2} = 0 \rightarrow x = \pm \sqrt{1/3}$$

2ND) WEIGHTS  $\left\{ \begin{array}{l} \text{Integrate Lagrange based function} \\ \text{Indet. obs} \end{array} \right.$

EXACT INT. OF

$x^0 = 1$	$\int_{-1}^1 dx = 2 = W_0 \cdot 1 + W_1 \cdot 1$	} WE ONLY NEED THIS 2 IN THIS CASE
$x^1 = x$	$\int_{-1}^1 x dx = 0 = W_0 \cdot (-1/\sqrt{3}) + W_1 \cdot (1/\sqrt{3})$	
$\vdots$	$\vdots$	

GL2 (ERROR TERM OF GAUSS-LEGENDRE RULE OF 2 NODES)

\*  $E = k \cdot f^{(M)}(\xi)$  f.s.  $\xi \in (-1,1)$  if  $f \in C^M([-1,1])$

$N = 2n+1 = 3 = 2n-1$

$f(x) = x^4 \rightarrow f^{(4)} = 4! = 24$

EXACT INT. OF:  $x^0 \rightarrow \int_{-1}^1 1 dx = 2 = W_0 + W_1$

$x^1 \rightarrow \int_{-1}^1 x dx = 0 = -\frac{1}{\sqrt{3}} W_0 + \frac{1}{\sqrt{3}} W_1$

$x^2 \rightarrow \int_{-1}^1 x^2 dx = \frac{2}{3} = Q$

$x^3 \rightarrow \int_{-1}^1 x^3 dx = \frac{2}{5} = Q + E = \left(-\frac{1}{\sqrt{3}}\right)^4 \cdot \frac{1}{3} + \left(\frac{1}{\sqrt{3}}\right)^4 \cdot \frac{1}{3} + k \cdot 24$

$[-1,1] \rightarrow [a,b]$

$x = \frac{a+b}{2} + \frac{b-a}{2} \xi$   
 $J = dx/d\xi$

\*  $\int_a^b f(x) dx = \int_{-1}^1 \underbrace{f\left(\frac{a+b}{2} + \frac{b-a}{2} \xi\right)}_{g(\xi)} \left(\frac{b-a}{2}\right) d\xi \approx Q = \frac{5}{9} g(-\sqrt{3}/5) + \frac{8}{9} g(0) + \frac{5}{9} g(\sqrt{3}/5) =$

$= \underbrace{\frac{5}{9} \cdot \frac{b-a}{2}}_{W_0} \cdot f\left(\underbrace{\frac{a+b}{2} + \frac{b-a}{2} \left(-\frac{\sqrt{3}}{5}\right)}_{x_0}\right) + \underbrace{\frac{8}{9} \cdot \frac{b-a}{2}}_{W_1} \cdot f\left(\underbrace{\frac{a+b}{2}}_{x_1}\right) + \underbrace{\frac{5}{9} \cdot \frac{b-a}{2}}_{W_2} \cdot f\left(\underbrace{\frac{a+b}{2} + \frac{b-a}{2} \left(\frac{\sqrt{3}}{5}\right)}_{x_2}\right)$

COMPOUND GAUSS (IN THE EXERCISES)

TERMINATION CRITERIUM

"Find  $\int_a^b f(x) dx$  with "error" less than ..."  
 ↳ TERMINATION CRITERIUM

GAUSS - CHEBYSHEV FORMULAS OF THE 1ST KIND (RULES)

$[a,b] = [-1,1]$  w  $w(x) = 1/\sqrt{1-x^2}$

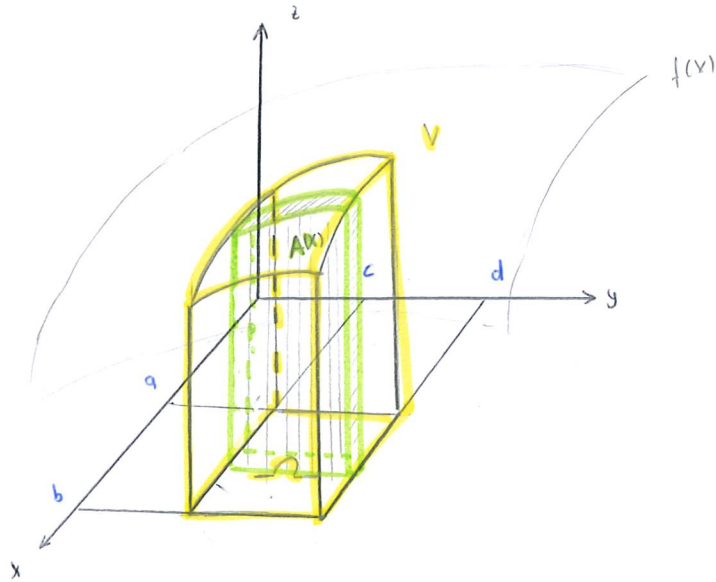
$\int_{-1}^1 f(x) \cdot \frac{1}{\sqrt{1-x^2}} dx \rightarrow$  nodes: the same as the ones of CH.  
 $T_{n+1}(x)$   $\{x_0, x_1, x_2, \dots, x_n\}$   
 n+1 NODES

↳ weights: they are equal and their sum is  $\pi$ ;  $w_i = \frac{\pi}{n+1} \rightarrow \int_{-1}^1 \frac{1}{\sqrt{1-x^2}} dx = \pi = \sum w_i$

GAUSS-LAGUERRE RULES

$$\int_0^{\infty} f(x) e^{-x} dx$$

$n$	$x_i$	$w_i$
0	$x$	$x$
1	$\frac{x}{2}$	$\frac{x}{2}$
2	$\left\{ \frac{x}{3}, \frac{x}{3}, \frac{x}{3} \right\}$	$\left\{ \frac{x}{3}, \frac{x}{3}, \frac{x}{3} \right\}$

NUMERICAL APPROX. OF SOME DOUBLE INT.

$$I \approx V = \int_a^b f(x, y) dx dy = \int_a^b A(x) dx$$

$$A(x) = \int_c^d f(x, y) dy$$

EXACT WHEN :  $N$  of  $f(x, y) \leq N$  of the RULE WE ARE USING

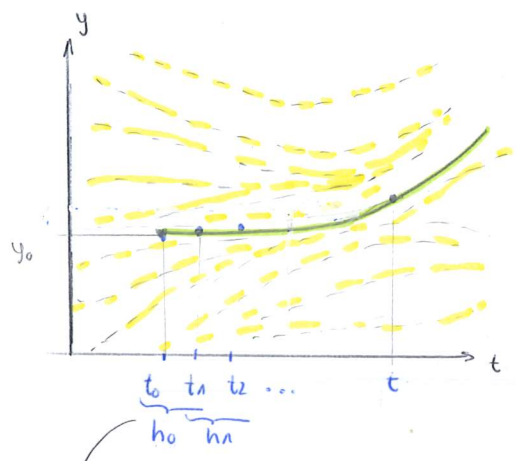


# CH. 4. INITIAL VALUE PROBLEMS

•  $t \equiv$  independent variable (time)

•  $y' = f(t, y)$  (simplest  $\frac{\text{ORD. DIFF. EQ}}{\text{EDA}}$ )

• slope field (campo de pendientes)  $\rightarrow$



ORD. DIFF. EQ

IVP  $\left\{ \begin{array}{l} \text{ODE } y' = f(t, y) \\ \text{IC } y(t_0) = y_0 \end{array} \right.$

INIT. COND.

\*  $\frac{y_1 - y_0}{h_0} = f(t_0, y_0) ; h_0 = t_1 - t_0$

$y_{k+1} = y_k + f(t_k, y_k) h_k$  EULER'S METHOD (E)

\*  $y^{(n)} = f(t, y, y', y'', \dots, y^{(n-1)})$

$u_1, u_2, u_3, \dots, u_n$

$$\begin{cases} u_1' = y' = u_2 \\ u_2' = u_3 = y'' \\ \vdots \\ u_{n-1}' = u_n = y^{(n-1)} \\ u_n' = f(t, u_1, u_2, \dots, u_n) = y^{(n)} \end{cases}$$

## NOTATION FOR SYSTEMS OF ODE

s. ODE-s  $\left\{ \begin{array}{l} y_1' = f_1(t, y_1, y_2, \dots, y_n) \\ y_2' = f_2(t, y_1, y_2, \dots, y_n) \\ \vdots \\ y_n' = f_n(t, y_1, y_2, \dots, y_n) \end{array} \right.$

ICs  $\left\{ \begin{array}{l} y_1(t_0) = y_{10} \\ y_2(t_0) = y_{20} \\ \vdots \\ y_n(t_0) = y_{n0} \end{array} \right.$

INIT. COND

\* UNKNOWNS:  $y_1(t), y_2(t), \dots, y_n(t)$

$$y' = f(t, y) \rightarrow \begin{cases} y_1' = f_1(t, y) \\ y_2' = f_2(t, y) \\ \vdots \\ y_n' = f_n(t, y) \end{cases} \xrightarrow{\text{(more compact)}} \boxed{y' = f(t, y)} \quad \text{s. ODE-s (COMPACT WAY)}$$

$y = (y_1, y_2, \dots, y_n)$

IC-s:  $\boxed{y(t_0) = y_0}$  (COMPACT WAY)

### LINEAR SYSTEM OF ODE-s (WITH CONSTANT COEFF.)

$$\begin{cases} y_1' = c_{11}(t)y_1 + c_{12}(t)y_2 + \dots + c_{1n}(t)y_n + g_1(t) = f_1(t, y) \\ y_2' = c_{21}(t)y_1 + c_{22}(t)y_2 + \dots + c_{2n}(t)y_n + g_2(t) = f_2(t, y) \\ \vdots \\ y_n' = c_{n1}(t)y_1 + c_{n2}(t)y_2 + \dots + c_{nn}(t)y_n + g_n(t) = f_n(t, y) \end{cases}$$

MATRIX FORM: 
$$y' = \underbrace{\begin{pmatrix} c_{11}(t) & c_{12}(t) & \dots & c_{1n}(t) \\ c_{21}(t) & c_{22}(t) & \dots & c_{2n}(t) \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1}(t) & c_{n2}(t) & \dots & c_{nn}(t) \end{pmatrix}}_{J(t) \text{ (JACOBIAN MATRIX)}} y + g(t) \rightarrow \boxed{y' = J(t)y + g(t)}$$

COEFFS DEPEND ON  $t$

\* LINEAR SYSTEM WITH constant coeffs:  $\boxed{y' = J \cdot y + g(t)}$

$J$  DOESN'T DEPEND ON  $t$

### JACOBIAN MATRIX OF A SYSTEM OF ODE-s

\* DEFINITION:  $J = \left( \frac{\partial f_i}{\partial y_j} \right)$  at  $(t, y)$

\* For a lin. system of ODE-s  $J(t) = \begin{pmatrix} c_{11}(t) & c_{12}(t) & \dots & c_{1n}(t) \\ c_{21}(t) & c_{22}(t) & \dots & c_{2n}(t) \\ \vdots & \vdots & \ddots & \vdots \\ c_{n1}(t) & c_{n2}(t) & \dots & c_{nn}(t) \end{pmatrix}$



PICARD'S THEOREM OF EXISTENCE AND UNIQUENESS TO INIT. VALUE PROBL. (IVP)

\* The IVP  $\begin{cases} y' = f(t, y) \\ y(t_0) = y_0 \end{cases}$  has a unique solution if  $f, \frac{\partial f}{\partial y} = f_y$  exist in

$R = \{ (t, y) \mid (t_0 < t < t_1) \wedge (\infty < y < \infty) \}$  and  $\exists L \in \mathbb{R}^+ / \forall (t, y_1), (t, y_2) \in R, |f(t, y_1) - f(t, y_2)| \leq L |y_1 - y_2|$

(REGION)

some constant

$L = \text{LIPSCHITZ} \rightarrow f$  is Lipschitzian (si cumple la DE ARRIA)

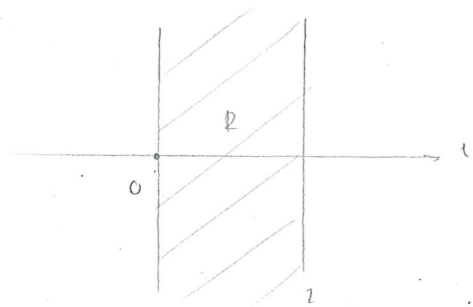
\*  $\exists f, f_y \exists$  and  $f_y$  is continuous in  $R \Rightarrow f$  is Lipschitzian

31/03/21

EXC. 1

$\exists!$  sol of  $\begin{cases} y' = 1 + t \sin(ty) \\ y(0) = y_0 \end{cases} \quad t \in [0, 2]$

$f(t, y)$



$$f(t, y) = 1 + t \sin(ty) \in C(R)$$

$$\frac{\partial f}{\partial y} = f_y(t, y) = t^2 \cos(ty) \in C(R)$$

$\in [0, 4] \quad \in [-1, 1]$

$f_y \in [-4, 4]$  bounded  $\Rightarrow f$  is Lipschitzian ( $L=4$ )

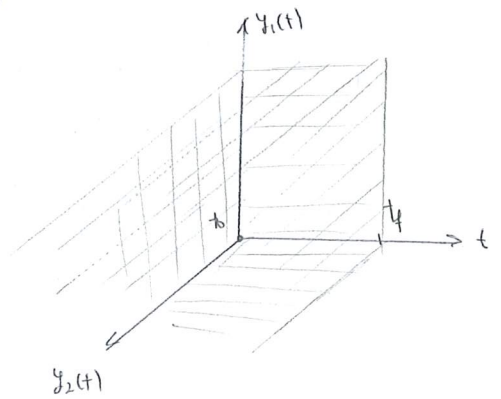
THEOREM OF  $\exists!$  OF SOL TO IVP (SYSTEMS) (Picard)

THE REGION IS A VOLUME

\* Let IVP be  $y' = f(t, y)$  with  $y(t_0) = y_0$  ( $t_0 \leq t \leq t_f$ ).

\* Let  $R = \{(t, y) / t_0 \leq t \leq t_f, y \in \mathbb{R}^n\} \in \mathbb{R}^{n+1}$

$$y(t) = \begin{pmatrix} y_1(t) \\ y_2(t) \\ \vdots \\ y_n(t) \end{pmatrix}$$



\* If  $f \in C(R)$  and  $\exists L \in \mathbb{R}^+$

$$\forall (t, y_1), (t, y_2), \|f(t, y_1) - f(t, y_2)\| \leq L \|(t, y_1) - (t, y_2)\| \quad (\text{with } \|\cdot\| \text{ any norm in } \mathbb{R}^{n+1}),$$

then IVP has a unique solution.  $\rightarrow f(t, y)$  is Lipschitzian.

THEOREM: If  $f_{y_1}, f_{y_2}, \dots, f_{y_n}$  are continuous and bounded in  $R$ , then  $f(t, y)$  is Lipschitzian.

Exc. 3

APPLY EULER'S METHOD, 5 steps with  $h=0.1$

$$\text{to solve } \begin{cases} y' = (1+t)y^2/2 = f(t, y) \\ y(0) = 1 \end{cases}$$

$$f = f(t, y) = (1+t) * y^2 / 2 ;$$

$$t_0 = 0 ; y_0 = 1 ; h = 0.1 ;$$

$$y_{k+1} = y_k + f(t_k, y_k) * h_k$$

$$y_1 = y_0 + f(t_0, y_0) * h \quad (= 1.05)$$

$$t_1 = t_0 + h ;$$

$$y_2 = y_1 + f(t_1, y_1) * h \quad (= 1.1106375)$$

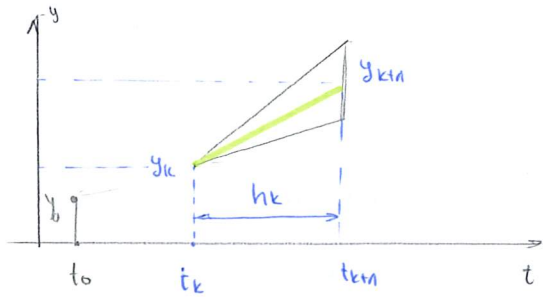
$$t_2 = t_1 + h ; t_3 = t_2 + h ; t_4 = t_3 + h ; t_5 = t_4 + h$$

$$y_3 = y_2 + f(t_2, y_2) * h ;$$

$$y_4 = y_3 + f(t_3, y_3) * h ;$$

$$y_5 = y_4 + f(t_4, y_4) * h ;$$

ENHANCED EULER'S (HENN'S) METHOD (E)



$$y_{k+1} = y_k + f(t_k, y_k) \cdot h_k \quad (\text{EULER})$$

AVERAGE SLOPE

$$y_{k+1} = y_k + \frac{f(t_k, y_k) + f(t_k + h_k, y_k + f(t_k, y_k) \cdot h_k)}{2} \cdot h_k \quad (\text{ENHANCED EULER})$$

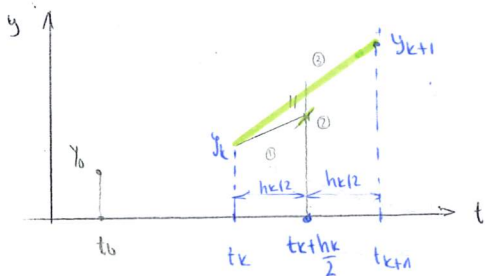
$$k_1 = f(t_k, y_k) h_k$$

$$k_2 = f(t_k + h_k, y_k + k_1) h_k$$

$$y_{k+1} = y_k + \frac{k_1 + k_2}{2}$$

(RUNGE-KUTTA STYLE)  
(RK 2)

MODIFIED EULER METHOD (MIDPOINT) (E)



$$y_{k+1} = y_k + f(t_k + h_k/2, y_k + f(t_k, y_k) h_k/2) \cdot h_k$$

(ADVANCED FORMULA FOR MIDPOINT METHOD)

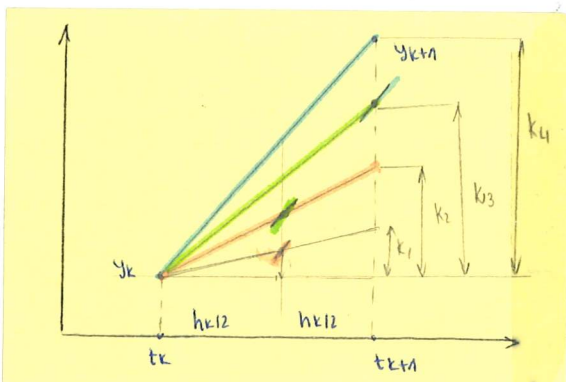
$$k_1 = f(t_k, y_k) h_k$$

$$k_2 = f(t_k + h_k/2, y_k + \frac{k_1}{2}) h_k$$

$$y_{k+1} = y_k + k_2$$

(RUNGE-KUTTA STYLE)

RUNGE-KUTTA METHOD OF ORDER 4 (RK4)



$$k_1 = f(t_k, y_k) h_k$$

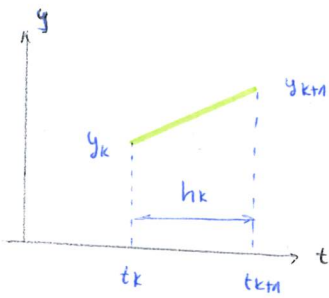
$$k_2 = f(t_k + h_k/2, y_k + k_1/2) h_k$$

$$k_3 = f(t_k + h_k/2, y_k + k_2/2) h_k$$

$$k_4 = f(t_k + h_k, y_k + k_2) h_k$$

$$y_{k+1} = y_k + \frac{k_1 + 2k_2 + 2k_3 + k_4}{6}$$

BACKWARD EULER METHOD (I)



$$y_{k+1} = y_k + f(t_{k+1}, y_{k+1}) h_k$$

unknown on both sides → implicit

\* unconditionally stable (ADVANTAGE)

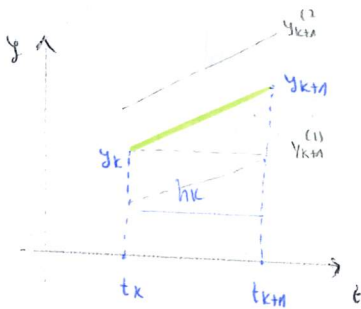
\* computational cost (DISADVANTAGE)

IMPLICIT VS. EXPLICIT METHOD

(I) unknown in both sides → implicit: we will iterate (↑↑ comp. cost BUT ↑↑ stability)

(E) we can obtain it directly with the data we have → explicit (EULER)

TRAPEZOIDAL METHOD (I)



$$y_{k+1} = y_k + \frac{f(t_k, y_k) + f(t_{k+1}, y_{k+1})}{2} h_k$$

∞ comput. cost  
OC = 2 (ORDER OF CONV.) → OPTIMAL

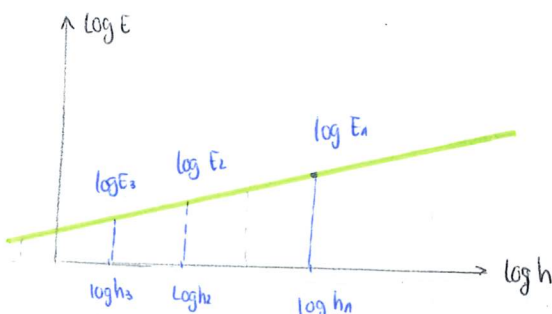
ORDER OF CONVERGENCE / ORDER OF PRECISION: (I)

\* COMPUTER EXAM

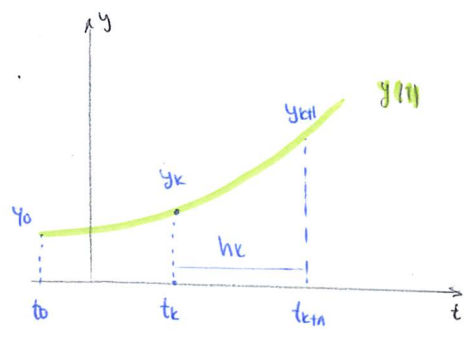
$E \approx k \cdot h^p$  (ERROR)

CON. COEFF.  $\log = \ln$  ( $\log_{10} = \log$ )

$\log(E) = \log(k h^p) = \log(k) + p \log(h)$



TAYLOR BASED METHODS (E)



$$y(t_{k+h}) = y(t_k) + y'(t_k)h + \frac{y''(t_k)h^2}{2!} + \frac{y'''(t_k)h^3}{3!} + \dots$$

(TAYLOR'S EXPANSION)

$$y' = f(t, y)$$

$$y'' = \frac{d^2y}{dt^2} = \frac{\partial f}{\partial t} + \frac{\partial f}{\partial y} \cdot \frac{dy}{dt} = f_t + f_y \cdot f$$

...

(NOT ALWAYS USEFUL) (OC=2)

$$y_{k+h} = y_k + \frac{f(t_k, y_k)h + f_t(t_k, y_k) + f_y(t_k, y_k) \cdot f(t_k, y_k)}{2!} h^2 + \dots$$

CHARACTERISTICS OF RUNGE-KUTTA METHOD

- \* explicit
- \* orders: 1, 2, 3, ..., 40, ..., 20, ...  
(n.e. f/st)
- \* computational cost: number of evaluations of f per step = order until order 4  
for ↑ orders: n.e.f/st > ORDER
- \* "popular"
- \* NOT TOO good for "stiff" systems of ODE's  
↳ RÍGIDO

EXC. 6 APPLY THE MODIFIED EULER METHOD: h=0.1

$$\begin{cases} y' = -y + t + 1 \\ y(0) = 1 \end{cases} \quad t \in [0, 1]$$

f = @(t,y) - y + t + 1;

t0 = 0; h = 0.1; t1 = h; t2 = 2\*h; t3 = 3\*h;

y0 = 1

y1 = y0 + f(t0 + h/2, y0 + f(t0, y0) \* h/2) \* h

y2 = y1 + f(t1 + h/2, y1 + f(t1, y1) \* h/2) \* h

...

EXC. 7 RK4 :  $\begin{cases} y' = t+y & t \in [0,1] \\ y(0) = 0 & h = 0.1 \end{cases}$

$f = @ (t,y) \quad t+y ; \quad h = 0.1;$

$t_0 = 0 ; \quad t_1 = h, \quad t_2 = 2*h ; \quad \dots ; \quad t_{10} = 10*h ;$

$y_0 = 0 ;$

$k_1 = f(t_0, y_0) * h ;$

$k_2 = f(t_0 + h/2, y_0 + k_1/2) * h ;$

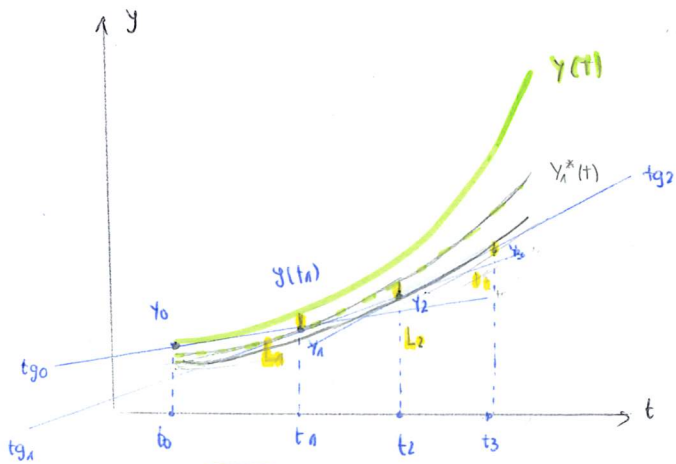
$k_3 = f(t_0 + h/2, y_0 + k_2/2) * h ;$

$k_4 = f(t_0 + h, y_0 + k_3) * h ;$

$y_1 = y_0 + (k_1 + 2*k_2 + 2*k_3 + k_4) / 6 ;$

↳ OTRA VEZ PARA (1) :  $k_1 = f(t_1, y_1) \dots$

STABILITY OF ODES



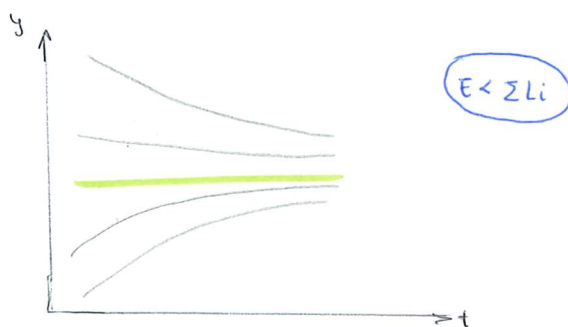
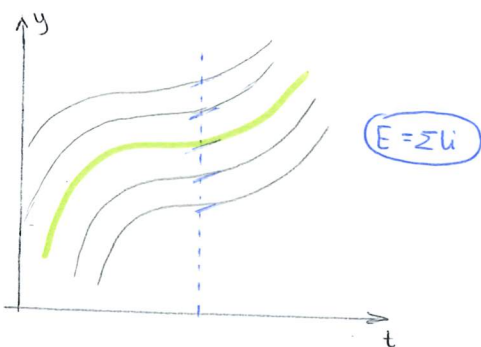
$E > \sum L_i$  (HACIENDO tg A LA CURVA POR  $y_0 \dots$ )

\*  $L_i \equiv$  LOCAL ERROR : attributable to step i

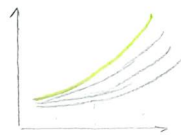
$E > L_1 + L_2 + L_3 + \dots$  (IN OUR DRAWING)

$$E_k \geq \sum_{i=1}^k L_i$$

GREATER, EQUAL OR LOWER



UNSTABLE ODE :  $E > \sum Li \rightarrow \frac{\partial f}{\partial y} > 0$



$$J = \left( \frac{\partial f}{\partial y_i} \right)$$

NEUTRAL ODE :  $E = \sum Li \rightarrow \frac{\partial f}{\partial y} = 0$



\* SINGLE ODE :  $J = \frac{\partial f}{\partial y} = f_y$

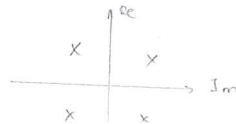
STABLE ODE :  $E < \sum Li \rightarrow \frac{\partial f}{\partial y} < 0$



\* THE STABILITY DEPENDS ON THE ODE ITSELF, NOT IN THE SOLVING METHOD

STABILITY OF SYSTEMS OF ODES

UNSTABLE S.ODEs :

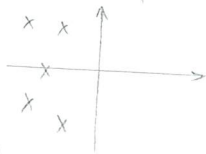


$\lambda =$  EIGENVALUES OF THE S.ODE

NEUTRAL S.ODEs :



STABLE S.ODEs :

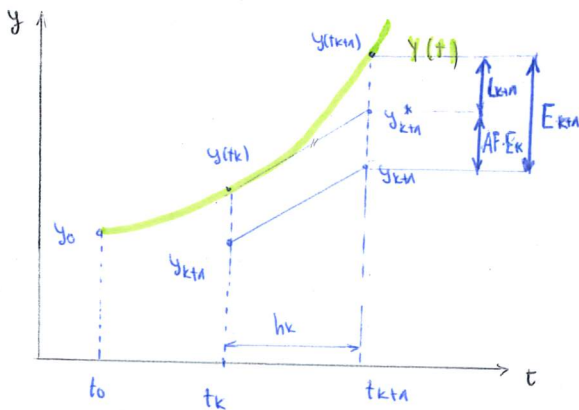


RIGID

\* A sys of ODEs is STIFF if it is a system of ODEs and it is stable and the real and/or imaginary parts of the  $\text{eig}(J)$  are "very" disproportionate

↳ WE CALCULATE THE  $\text{eig}(J)$  AND CHECK THE STIFFNESS

TRUNCATION ERRORS (EULER'S METHOD)



\*  $y_{k+1}^*$  = point of arrival with Euler's method if you start from  $(t_k, y(t_k))$

EULER:  $y_{k+1} = y_k + f(t_k, y_k) h_k$   
 $y_{k+1}^* = y(t_k) + f(t_k, y(t_k)) h_k$

$f \in C^2$

LOCAL ERROR:

$$L_{k+1} = [y(t_{k+1})] - [y_{k+1}^*] = [y(t_k) + y'(t_k)h_k + \frac{y''(\xi)}{2!} h_k^2] - [y(t_k) + f(t_k, y(t_k)) h_k] = \frac{y''(\xi) h_k^2}{2!}$$

↳  $L_{k+1} = O(h_k^2)$

$$\boxed{OC = LOC - 1}$$

(ORD. OF CONV. = LOCAL ORD. CONV. - 1)

Euler method is of order 1

AMPLIFICATION FACTOR OF EULER'S METHOD (AF)

GLOBAL ERROR :

$$E_{k+1} = y(t_{k+1}) - y_{k+1} = \underbrace{y(t_{k+1}) - y_{k+1}^*}_{L_{k+1}} + \underbrace{(y_{k+1}^* - y_{k+1})}_{\text{WE ADD } (1)^{k-1}}$$

$$= L_{k+1} + (y(t_k) + f(t_k, y(t_k))h_k) - (y_k + f(t_k, y_k)h_k) = L_{k+1} + E_k + (f(t_k, y(t_k)) - f(t_k, y_k))h_k =$$

$$= L_{k+1} + E_k + f_y(t_k, \xi) E_k \cdot h_k = \boxed{L_{k+1} + E_k (1 + J(t_k, \xi) h_k)}$$

AF

\*  $E_{k+1} = L_{k+1} + AF \cdot E_k \rightarrow \boxed{AF = (1 + J(t_k, \xi) h_k)}$

- |  $|AF| = 1 \rightarrow E = \sum L_i$  (NEUTRAL)
- |  $|AF| < 1 \rightarrow E < \sum L_i$  (STABLE)
- |  $|AF| > 1 \rightarrow E > \sum L_i$  (UNSTABLE)

20/04/21

\* Absolute stability condition of Euler method:  $|AF| < 1$

$$-1 < AF < 1$$

$$-1 < 1 + J \cdot h < 1$$

$$-2 < Jh < 0$$

$|AF| < 1 \iff$  METHOD STABLE

$$\boxed{Jh \in (-2, 0)}$$

STABILITY INTERVAL

If ODE is unstable :  $J > 0 \rightarrow J \cdot h > 0 \rightarrow Jh \notin (-2, 0) \rightarrow$  METHOD UNSTABLE

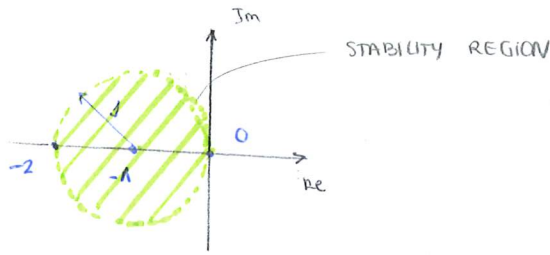
If ODE is stable :  $J < 0 \rightarrow -2 < Jh < 0 \rightarrow \boxed{h < \frac{-2}{J} = h_t}$  : threshold step size for absolute stability of the method.



STABILITY OF EULER'S METHOD WITH SODEs

Euler's method will be stable if and only if :  $eigs(Jh) = h \cdot eigs(J)$

(AS IN LIN. ALGEBRA)

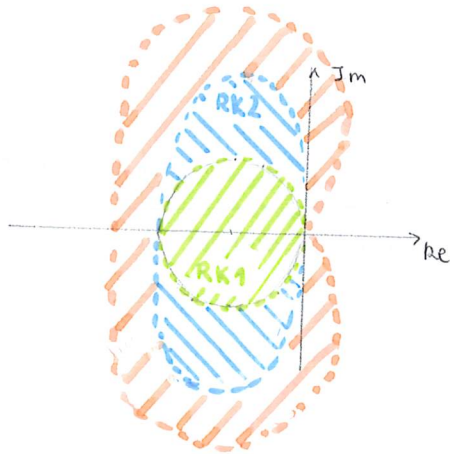


ABSOLUTE STABILITY INTERVAL OF ANY METHOD

By definition, it is the interval where  $J \cdot h$  must lie for the method to be stable solving an ODE.

ABS. STAB. REGION OF ANY METHOD

Def: The region of the complex plane where  $eigs(Jh)$  must lie for the method to be stable solving a SODEs.



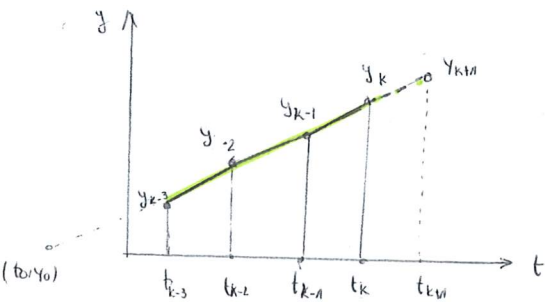
\* They are symmetric with respect to the Re axis

\* solving system of real ODEs

$$\begin{cases} J = (f_y); eigs(J) = f_y \\ eigs(Jh) = h \cdot f_y \end{cases}$$

(E) MULTISTEP METHODS (LMM) (LINEAR MULTISTEP METHOD)

AB4

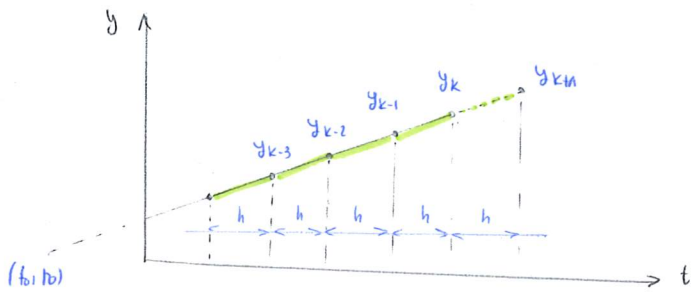


• ABn = Adams Bashforth methods are explicit multistep methods

• Multistep methods = methods "with memory"  
• single-step methods = methods "without memory"

LAST VALUE OBTAINED (LAST STEP)

$y' = f(t, y)$



$$\begin{cases} y' = f(t, y) \\ y_i = f(t_i, y_i) \end{cases}$$

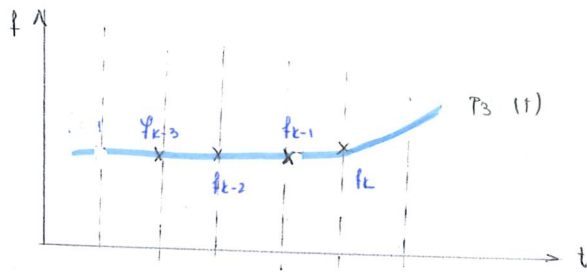
$$y(t_{kM}) = y(t_k) + \int_{t_k}^{t_{kM}} y'(t) dt \quad \text{(BURROWS LAW) (EXACT)}$$

$$y_{kM} = y_k = \int_{t_k}^{t_{kM}} f(t) dt$$

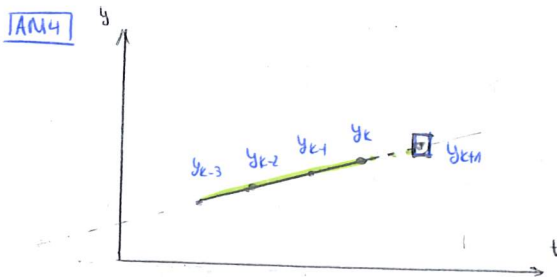
DISCRETE

$$y_{kM} = y_k + \frac{h}{24} (55f_k - 59f_{k-1} + 37f_{k-2} - 9f_{k-3})$$

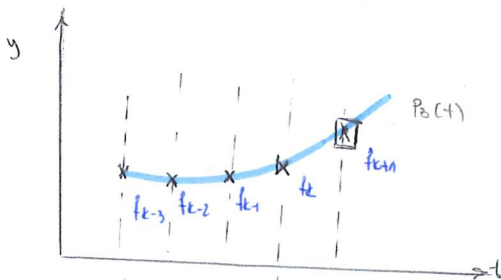
(AB4 advanced formula)



\* TO calculate  $y_{kM}$  we are not increasing the computational cost  
 ↳ not extra



• AM = Adams-Moulton method of order n (AMn) is an IMPPLICIT multistep method.

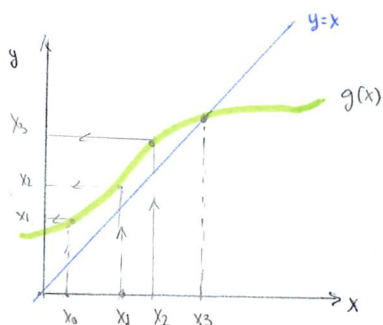


$$y_{kM} = y_k + \frac{h}{24} (9f(t_{kM}, y_{kM}) + 19f_k - 5f_{k-1} + f_{k-2})$$

(AM4 advanced formula)

\* comp. cost: ∞

FIXED POINTS INTER. (rev)



$$\begin{cases} y_{k+1} = g(y_{k+1}) \\ x = g(x) \end{cases}$$

cond. for convergence  
 $|g'(x)| < 1$  (and  $x \in$  to that neighbourhood)

we suppose:  $x_0 \rightarrow x_1 = g(x_0)$   
 $x_2 = g(x_1)$   
 $x_3 = g(x_2)$   
 $\vdots$   
 $x_{k+1} = g(x_k)$  } IT CONVERGES

EXC 14

$x_i$	$y_i$	$f_i$
0	0	0
0.2	0.02103	0.22103
0.4	0.09137	0.49137
0.6	0.22156	0.82156

4 points  $\rightarrow$  METHOD OF ORDER 4

Apply Adams PC method of order 4 for 2 more

Points of:  $\begin{cases} y' = x+y \\ y(0) = 0 \end{cases}$  with  $h=0.2$ ,Scheme PC  $(2 \cdot 10^{-3})$  (Rounding 5 dec)

(NB) AB4:  $y_{nm} = y_n + \frac{h}{24} (55f_n - 59f_{n-1} + 37f_{n-2} - 9f_{n-3})$  pred.

AM4:  $y_{nm} = y_n + \frac{h}{24} (9f_{n+1} + 19f_n - 5f_{n-1} + f_{n-2})$  "corrected"

WITH CORRECT NOTATION:

$$f = @ (t, y) t + y;$$

$$y_0 = 0;$$

$$h = 0.2;$$

$$t_0 = 0; t_1 = h; t_2 = 2 \cdot h; t_3 = 3 \cdot h; t_4 = 4 \cdot h; t_5 = 5 \cdot h;$$

$$y_1 = 0.02103; y_2 = 0.09137; y_3 = 0.22156;$$

$(t_0 = \dots; f_1 = \dots) \rightarrow$  we don't really need it, we already have  $f(t, y)$ ,  $t$  and  $y$

$$y_{4p} = y_3 + h/24 * (55 * f(t_3, y_3) - 59 * f(t_2, y_2) + 37 * f(t_1, y_1) - 9 * f(t_0, y_0)); \rightarrow 0.42469$$

$$y_{4c1} = y_3 + h/24 * (9 * f(t_4, y_{4p}) - 19 * f(t_3, y_3) - 5 * f(t_2, y_2) + f(t_1, y_1)); \rightarrow 0.42486$$

$$t_{c1} = y_{4c1} - y_{4p}; \rightarrow 0.00019 > 2 \cdot 10^{-5} \text{ we have to keep on iterating}$$

$$y_{4c2} = y_3 + h/24 * (9 * f(t_4, y_{4c1}) - 19 * f(t_3, y_3) - 5 * f(t_2, y_2) + f(t_1, y_1)); \rightarrow 0.42487$$

$$t_{c2} = y_{4c2} - y_{4c1}; \rightarrow 0.00001 < 2 \cdot 10^{-5} \text{ stop iterating}$$

$$y_4 = y_{4c2};$$

(same for  $y_5$ )

Exc. 13

solve  $\begin{cases} y' = (1+x) \cdot y^2/2 \\ y(0) = 1 \end{cases} \quad x \in [0, 0.7]$

$h = 0.1$ ;  $\rightarrow$  Milne's PC method with RK4

$$\begin{cases} y(0.1) \approx y_1 = 1.055409 \\ y(0.2) \approx y_2 = 1.123596 \\ y(0.3) \approx y_3 = 1.208459 \end{cases}$$

$$\begin{cases} y_{n+1} = y_{n-3} + \frac{4h}{3} (2f_n - f_{n-1} + 2f_{n-2}) & \text{MILNE} \\ y_{nm} = y_{n-1} + \frac{h}{3} (f_{nm} + 4f_n + f_{n-1}) & \text{SIMPSON} \end{cases}$$

both order 4 with scheme P(EC)E PC<sup>4</sup>

coding:

$f = @(t, y) (1+t) * y^2 / 2;$

$h = 0.1;$

$t_0 = 0; t_1 = h; t_2 = 2 * h; t_3 = 3 * h; \dots; t_7 = 7 * h;$

$y_0 = 1; y_1 = 1.055409; y_2 = 1.123596; y_3 = 1.208459;$

$y_{4P} = y_0 + 4 * h / 3 * (2 * f(t_3, y_3) - f(t_2, y_2) + 2 * f(t_1, y_1));$

$y_{4C1} = y_2 + h / 3 * (f(t_4, y_{4P}) + 4 * f(t_3, y_3) + f(t_2, y_2));$

$t_c = y_{4C1} - y_{4P}; \quad (\text{if } t_c < 1) \rightarrow y_4 = y_{4C1}$

CONVERGENCE AND STABILITY OF LINEAR MULTISTEP METHODS

0) Identify the number of steps of the method,  $k$ .

1) Identify advance formula with:

$$y_{n+k} = \sum_{j=0}^{k-1} \alpha_j y_{n+j} + h \sum_{j=0}^k \beta_j f_{n+j}$$

$\alpha_k > 1$  ALWAYS  
 $\uparrow$

2) Write characteristic polynomials:

$$p(z) = \sum_{j=0}^k \alpha_j z^j$$

$$\sigma(z) = \sum_{j=0}^k \beta_j z^j$$

3) Method is consistent iff:  $p(1) = 0; p'(1) = \sigma(1)$   
(CONSISTENCY)

$\hookrightarrow$  consistent means:  $\lim_{h \rightarrow 0} \frac{L_{k+1}}{h} \stackrel{\text{EULER}}{=} 0$

4) stable if:  $| \text{roots of } p(z) | \leq 1$  all (and if 1, simple roots)

5) convergence of order  $p$  if:

$$\frac{1}{m} \sum_{j=0}^k \alpha_j j^m = \sum_{j=0}^k \beta_j j^{m-1}$$

27104121

EXC. 10

a) study conv. and order of conv. of:

MILNE'S METHOD:  $y_{(n)} = y_{(n-3)} + \frac{4h}{3} (2f_n - f_{n-1} + 2f_{n+2})$

0) NUMBER OF STEPS :  $K = (n+1) - (n-3) = 4$

1) IDENTIFY ADV. FORMULA :  $y_{n+4} = \sum_{i=0}^3 \alpha_i y_{ni} + h \sum_{i=0}^4 \beta_i f_{ni}$

$$\begin{cases} y_{n+4} = -\alpha_0 y_n - \alpha_1 y_{n+1} - \alpha_2 y_{n+2} - \alpha_3 y_{n+3} + h(\beta_0 f_n + \beta_1 f_{n+1} + \beta_2 f_{n+2} + \beta_3 f_{n+3} + \beta_4 f_{n+4}) \\ y_{n+4} = y_n + \frac{4h}{3} (2f_{n+3} - f_{n+2} + 2f_{n+1}) \quad (\text{adaptation of Milne's method (13)}) \end{cases}$$

$$\begin{cases} \alpha_0 = -1 \\ \alpha_1 = \alpha_2 = \alpha_3 = 0 \\ \alpha_4 = 1 \end{cases} \quad \begin{cases} \beta_0 = \beta_4 = 0 \\ \beta_1 = \beta_3 = 8/3 \\ \beta_2 = -4/3 \end{cases} \quad \text{WHEN THE LAST } \beta = 0 \rightarrow \text{EXPLICIT METHOD}$$

2) CHARACT. POL :  $P(z) = -1 + z^4$        $D(z) = \frac{8}{3}z - \frac{4}{3}z^2 + \frac{8}{3}z^3$

3) CONSISTENCY :  $P(1) = -1 + 1 = 0 \quad \checkmark$        $P'(1) = 3 \cdot 1^3 = \frac{8}{3} \cdot 1 - \frac{4}{3} \cdot 1 + \frac{8}{3} \cdot 1^3 = 3 = D(1) \quad \checkmark$

4) STABILITY :  $P(z) = 0 \rightarrow z^4 = 1 \Rightarrow (z^2)^2 = 1 \begin{cases} z^2 = +1 \\ z^2 = -1 \end{cases} \begin{cases} z = \pm 1 \\ z = \pm i \end{cases}$

modulus of the roots  $\leq 1 = 1 \rightarrow$  all simple  $\checkmark$



5) CONV. OF ORDER  $p$ :  $\frac{1}{m} \sum_{j=0}^4 \alpha_j j^m = \sum_{j=0}^4 \beta_j j^{m-1} \quad (m=1, 2, \dots)$

$m=1 \rightarrow \frac{1}{1} (-1 \cdot 0^1 + \underbrace{1 \cdot 4^1}_4) = \frac{8}{3} \cdot 1^0 + \frac{-4}{3} \cdot 2^0 + \frac{8}{3} \cdot 3^0 \rightarrow 4 = 4$

$m=2 \rightarrow \frac{1}{2} (1 \cdot 4^2) \dots \rightarrow 8 = 8$

$m=3 \rightarrow \frac{1}{3} (1 \cdot 4^3) = \frac{8}{3} \cdot 1^2 - \frac{4}{3} \cdot 2^2 + \frac{8}{3} \cdot 3^2 \rightarrow \dots$

$$m=4 \rightarrow \dots$$

$$m=5 \rightarrow \text{DEGREE OF CONV} : \boxed{p=4}$$

$$c) \text{ AM4: } y_{n+1} = y_n + \frac{h}{24} (9f_{n+1} + 19f_n - 5f_{n-1} + f_{n-2})$$

$$d) \left( \nu = (n+1) - (n-2) = 3 \right)$$

$$1) y_{n+3} = \sum_{j=0}^2 \alpha_j y_{n+j} + h \sum_{j=0}^3 \beta_j f_{n+j} = (\alpha_0 y_n + \alpha_1 y_{n+1} + \alpha_2 y_{n+2}) + h(\beta_0 f_n + \beta_1 f_{n+1} + \beta_2 f_{n+2} + \beta_3 f_{n+3})$$

(+2)

$$y_{n+3} = y_{n+2} + \frac{h}{24} (9f_{n+3} + 19f_{n+2} - 5f_{n+1} + f_n)$$

$$\begin{cases} \alpha_0 = \alpha_1 = 0 \\ \alpha_2 = -1 \\ \alpha_3 = 1 \end{cases} \quad \begin{cases} \beta_0 = -1/24 & \beta_1 = 19/24 \\ \beta_2 = -5/24 & \beta_3 = 9/24 \end{cases}$$

$$2) \boxed{f(z) = -z^2 + z^3} \quad \boxed{\sigma(z) = \frac{z}{24} - \frac{5z^2}{24} + \frac{19z^3}{24} + \frac{9z^4}{24}}$$

$$3) p(1) = -1 + 1 = 0 \checkmark$$

$$p'(1) = -2 + 3 = 1 \checkmark$$

$$\sigma(1) = \frac{1}{24} - \frac{5}{24} + \frac{19}{24} + \frac{9}{24} = \frac{24}{24} = 1$$

CONSISTENT

$$4) p(z) = 0 \rightarrow z^2(z-1) = 0 \begin{cases} z=0 \text{ (DOUBLE)} \\ z=1 \end{cases} \rightarrow |z| \leq 1; \text{ 1 SIMPLE } \checkmark \rightarrow \text{STABLE}$$

$$5) \frac{1}{m} \sum_{j=0}^3 \alpha_j j^m = \sum_{j=0}^3 \beta_j j^{m-1} \rightarrow \frac{1}{m} (-2^m + 3^m) = \frac{-5}{24} \cdot 2^{m-1} + \frac{19}{24} 3^{m-1} + \frac{9}{24} 4^{m-1} + \left( \frac{1}{24} \cdot 0^{m-1} \right)$$

$$\textcircled{m=1} \rightarrow -2 + 3 = 1 = \frac{-5}{24} + \frac{19}{24} + \frac{9}{24} + \frac{1}{24} = 1 \checkmark \quad (1=1)$$

$$\textcircled{m=2} \rightarrow \frac{1}{2} (-4 + 9) = \frac{5}{2} = \frac{-5}{12} + \frac{19}{8} + \frac{9}{6} =$$

$\vdots$   $\neq$

$\rightarrow$  ORDER: p=4

$\textcircled{m=3}$   $\neq$

1.) Aux. pol.  $\Pi(z) = f(z) - \bar{h} \cdot \sigma(z)$  ( $f, \sigma = 1^{st}, 2^{nd}$  char. pol.)

stability region  $\{ \bar{h} / |z(\bar{h})| < 1 \}$

$\Pi(z) = 0 \rightarrow z = z(\bar{h}) \rightarrow |z(\bar{h})| < 1$

Ex.) a) stability region of Euler's M.

EULER:  $y_{n+1} = y_n + f(t_n, y_n) h$

$y_{n+1} = y_n + h f_n$

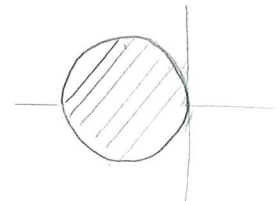
0)  $k = (n+1) - (n) = 1 \rightarrow 1) y_{n+1} = -\sum_{i=0}^0 \alpha_i y_{n+i} + h \sum_{i=0}^1 \beta_i f_{n+i}$

$$\left. \begin{aligned} y_{n+1} &= -\alpha_0 y_n + h(\beta_0 f_n + \beta_1 f_{n+1}) \\ y_{n+1} &= y_n + h f_n \end{aligned} \right\} \begin{cases} \alpha_0 = -1; \alpha_1 = 1 \\ \beta_0 = 1; \beta_1 = 0 \end{cases}$$

2)  $f(z) = -1 + z$        $\sigma(z) = 1$

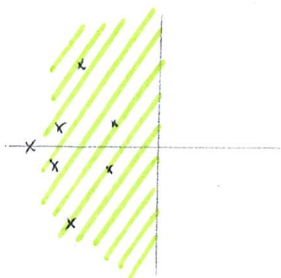
1)  $\Pi(z) = (-1 + z) - \bar{h} \cdot 1 = -1 + z - \bar{h} = 0 \rightarrow z(\bar{h}) = \bar{h} + 1$

$$\left\{ \begin{aligned} |1 + \bar{h}| &< 1 \\ |\bar{h} - (-1)| &< 1 \end{aligned} \right.$$



28/04/21

UNCONDITIONAL STABILITY (A-STABLE)



Backward Euler (AM1)  $\rightarrow$  order 1  
Trapezoidal method (AM)  $\rightarrow$  order 2

\* There's no a method with order  $> 2$  that is unconditionally stable.

## CONVERGENCE OF PC METHODS

Adams' Method of order 4

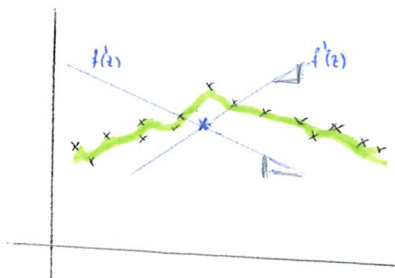
AB4 Expl. → To improve precision (corrections)  
AM4 Impl.

\* Each method has its corrections, not always of the same order of conv.



# CH. 3. NUMERICAL DIFFERENTIATION

$k =$  order of diff. of  $f$   $\rightarrow f^{(k)}(z)$   
 $z =$  point we are estimating



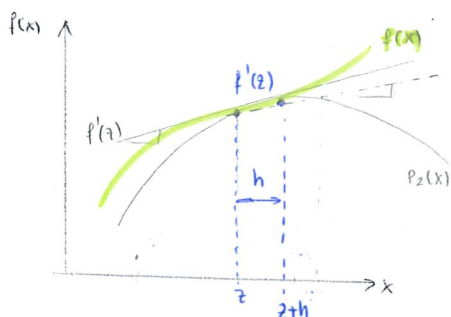
\* we can have noise

\* Then, we have more than one  $f'(z)$  because it's not exact.

$\rightarrow$  we are going to be more accurate when working with numerical differentiation

(\* we are going to calculate the approximate linear system.)

## INTERPOLATION IDEA

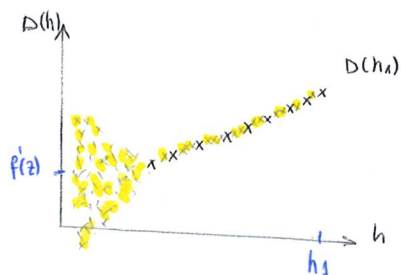


$k=1$

$$f'(z) = \lim_{h \rightarrow 0} \frac{f(z+h) - f(z)}{h}$$

$$f'(z) \approx \frac{f(z+h) - f(z)}{h} + O(h) \quad OC=1 \text{ (trunc.)}$$

$\uparrow$   
small  $h$   
D



• when  $h$  is very small

\* FORWARD DIFF. FORMULA:

$$f'(z) = \frac{f(z+h) - f(z)}{h} + O(h) \quad // \quad OC=1 \text{ (trunc.)}$$

$$f'(z) = \frac{f(z+h) - f(z-h)}{2h} + O(h^2) \quad // \quad OC=2$$

\* BACKWARD DIFF. FORMULA:

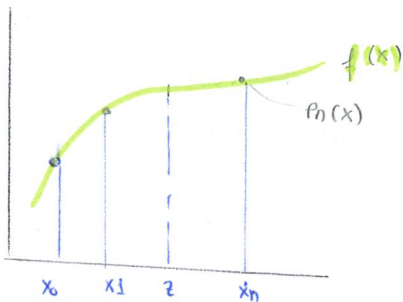
$$f'(z) = \frac{f(z) - f(z-h)}{h} + O(h) \quad // \quad OC=1$$

(K=2)

$$f''(z) \approx \frac{f'(z+h/2) - f'(z-h/2)}{h} \approx \frac{\frac{f(z+h) - f(z)}{h} - \frac{f(z) - f(z-h)}{h}}{h} = \frac{f(z+h) - 2f(z) + f(z-h)}{h^2}$$

$$f''(z) = \frac{f(z+h) - 2f(z) + f(z-h)}{h^2} + O(h^2) \quad // \quad OC=2$$

INTERPOLATORY NUMERICAL DIFFERENTIATION FORMULA



$$\begin{cases} f^{(k)}(z) \approx P_n^{(k)}(z) = D \\ f^{(k)}(z) = \underbrace{P_n^{(k)}(z)}_D + E \end{cases}$$

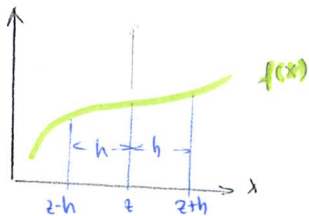
$$P_n(x) = L_0(x) f(x_0) + L_1(x) f(x_1) + \dots + L_n(x) f(x_n)$$

$$P_n^{(k)}(x) = L_0^{(k)}(x) f(x_0) + L_1^{(k)}(x) f(x_1) + \dots + L_n^{(k)}(x) f(x_n)$$

$$\begin{aligned} \overline{D} &= P_n^{(k)}(z) = L_0^{(k)}(z) f(x_0) + L_1^{(k)}(z) f(x_1) + \dots + L_n^{(k)}(z) f(x_n) = \\ &= A_0 f(x_0) + A_1 f(x_1) + \dots + A_n f(x_n) = \sum_{i=0}^n \underbrace{A_i}_{\text{COEFS // WEIGHTS}} f(x_i) \end{aligned}$$

↑  
NODES

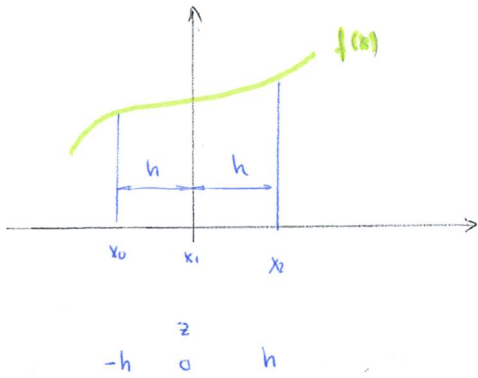
EXAMPLE: D FOR f'(z) USING z-h, z, z+h (CENTERED FORMULA)



$$\begin{aligned} P_2(x) &= \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} f(x_0) + \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} f(x_1) + \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)} f(x_2) = \\ &= \underbrace{\frac{(x-z)(x-z-h)}{-h(-2h)}}_{L_0(x)} f(x_0) + \underbrace{\frac{(x-z+h)(x-z-h)}{h(-h)}}_{L_1(x)} f(x_1) + \underbrace{\frac{(x-z+h)(x-z)}{2h \cdot h}}_{L_2(x)} f(x_2) \end{aligned}$$

$$\left. \begin{aligned} A_0 &= L_0'(z) ; & L_0'(z) &= \frac{(z-z) + (z-z-h)}{2h^2} \rightarrow A_0 = -1/2h \\ A_1 &= L_1'(z) ; & L_1'(z) &= \frac{(z-z+h) + (z-z-h)}{-h^2} \rightarrow A_1 = 0 \\ A_2 &= L_2'(z) ; & L_2'(z) &= \frac{(z-z+h) + (z-z)}{2h^2} \rightarrow A_2 = 1/2h \end{aligned} \right\} \boxed{D = \frac{f(z+h) - f(z-h)}{2h}}$$

EXAMPLE: D for  $f''(z)$  CENTERED, 3 NODES



$$\begin{aligned}
 f(x) &= 1 \rightarrow f''(0) = 0 = A_0 \cdot 1 + A_1 \cdot 1 + A_2 \cdot 1 \rightarrow A_0 + A_1 + A_2 = 0 \\
 f(x) &= x \rightarrow f''(0) = 0 = A_0(-h) + A_1 \cdot 0 + A_2 \cdot h = 0 \rightarrow A_0 = A_2 \\
 f(x) &= x^2 \rightarrow f''(0) = 2 = A_0(-h)^2 + A_1 \cdot 0 + A_2 \cdot h^2 \rightarrow A_0 + A_2 = 2/h^2 \\
 \rightarrow A_0 &= -3/h^2 = A_2 \rightarrow A_1 = -2/h^2
 \end{aligned}$$

$$D = \frac{1}{h^2} f(z-h) + \frac{(-2)}{h^2} f(z) + \frac{1}{h^2} f(z+h) = \frac{f(z-h) - 2f(z) + f(z+h)}{h^2}$$

TRUNCATION ERROR VIA INTERPOLATION

$$f^{(k)}(z) = \underbrace{P_n^{(k)}(z)}_D + E$$

$$\rightarrow E = f^{(k)}(z) - P_n^{(k)}(z) = (f(z) - P_n(z))^{(k)} = e^{(k)}(z) = (f[z_0, x_1, \dots, x_n, z] \Pi(z))^{(k)}$$

\*NEW THEOREM:

$$f[z_0, x_1, \dots, x_n, x]^{(k)} = k! f[z_0, x_1, \dots, x_n, \underbrace{x, x, \dots, x}_k] = \frac{k! f^{(n+k)}(\xi)}{(n+k)!}$$

f.s.  $\xi$  between the nodes and  $x$

CASUSTRY

• CASUSTRY:  $f'(z)$  ( $k=1$ )

$$\begin{aligned}
 E &= (f[z_0, x_1, x_2, \dots, x_{n+1}, z] \Pi(z))' = f[z_0, x_1, \dots, x_{n+1}, z] \Pi'(z) + f[z_0, \dots, x_{n+1}, z] \Pi'(z) \\
 &= \frac{1! f^{(n+2)}(\xi_1)}{(n+2)!} \Pi(z) + \frac{0! f^{(n+1)}(\xi_2)}{(n+1)!} \Pi(z) = \frac{f^{(n+2)}(\xi_1)}{(n+2)!} \Pi(z) + \frac{f^{(n+1)}(\xi_2)}{(n+1)!} \Pi(z)
 \end{aligned}$$

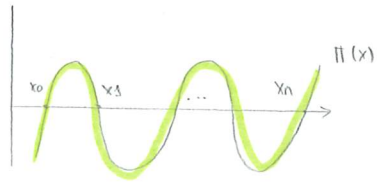
NEW THEOREM

$\xi_1, \xi_2$  between nodes and  $z$   
 $f \in C^{n+2}$

$$\Pi(x) = (x - x_0)(x - x_1) \dots (x - x_n)$$

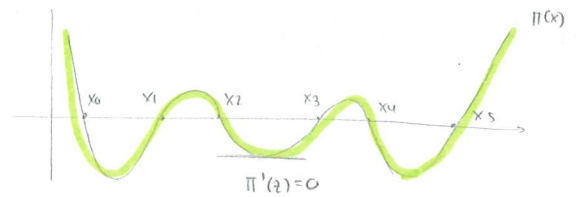
A)  $\Pi(z) = 0 \iff z$  is one of the nodes

$$E = \frac{f^{(n+1)}(\xi)}{(n+1)!} \Pi'(z)$$



$$\left\{ \begin{array}{l} f(x) = x^n \rightarrow f^{(n+1)}(x) = 0 \rightarrow E = 0 \\ f(x) = x^{n+1} \rightarrow f^{(n+1)}(x) = (n+1)! \neq 0 \rightarrow E \neq 0 \end{array} \right\} \boxed{N=n} \text{ if } z \text{ is a node to estimate } f'(z)$$

B)  $\Pi'(z) = 0 \iff$  Even number of nodes with central symmetry centered at  $z$ .



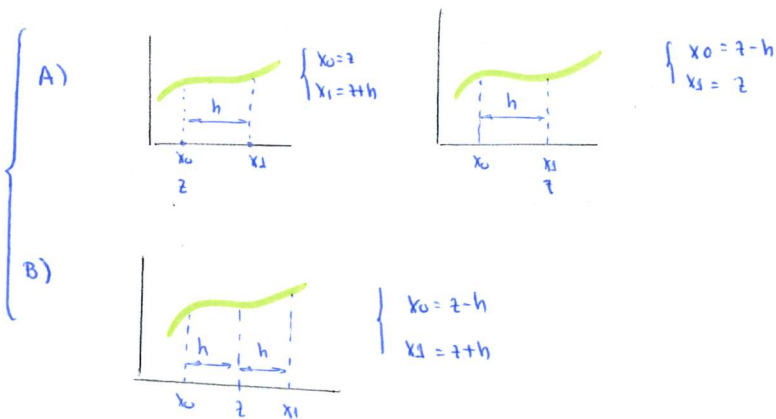
$$\left\{ \begin{array}{l} x^{n+1} \rightarrow E = 0 \\ x^{n+2} \rightarrow E \neq 0 \end{array} \right\} \boxed{N=n+1}$$

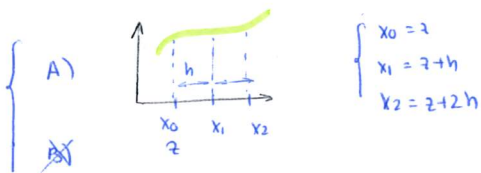
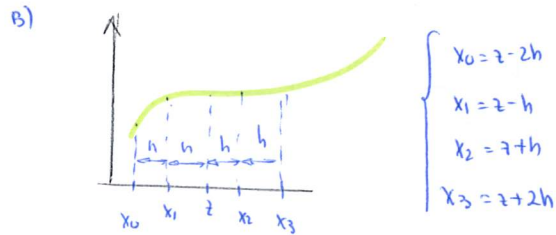
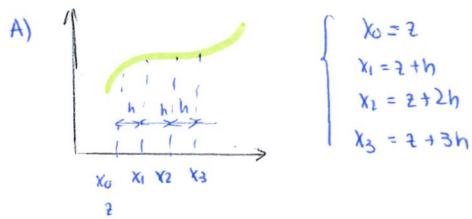
• CASUISTRY:  $f''(z)$  ( $k=2$ )

$$E = (f[x_0, \dots, x_n, z] \Pi(z))'' = (f[x_0, \dots, x_{n-1}, z] \Pi'(z) + f[x_0, \dots, x_n, z] \Pi'(z))' = \dots =$$

$$= 2 \frac{f^{(n+2)}(\xi_1)}{(n+3)!} \Pi(z) + 2 \frac{f^{(n+1)}(\xi_2)}{(n+2)!} \Pi'(z) + \frac{f^{(n+1)}(\xi_3)}{(n+1)!} \Pi''(z)$$

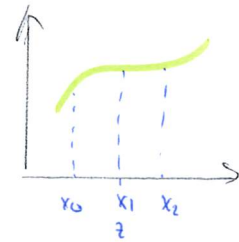
• CASES:  $f'(z)$  2 NODES



CASES  $f'(z)$  3 NODESCASES  $f'(z)$  4 NODESCASES  $f''(z)$  ( $k=2, n=2$ ) ( $nn=3$ )

$$P_2(x) = \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} f(x_0) + \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} f(x_1) + \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)} f(x_2) =$$

$$= \underbrace{\frac{(x-x_1)(x-x_2)}{-h(-2h)}}_{L_0(x)} f(x_0) + \underbrace{\frac{(x-z+h)(x-z-h)}{h(-h)}}_{L_1(x)} f(x_1) + \underbrace{\frac{(x-z+h)(x-z)}{2h \cdot h}}_{L_2(x)} f(x_2)$$



$$\begin{cases} \pi(x) = (x-x_0)(x-x_1)(x-x_2) = (x-z+h)(x-z)(x-z-h) \\ \pi'(x) = (x-z)(x-z-h) + (x-z+h)(x-z-h) + (x-z+h)(x-z) \\ \pi(z) = 0; \quad \pi'(z) = h^2; \quad \pi''(z) = 0 \end{cases}$$

$$\rightarrow \epsilon = 2 \cdot \frac{f''(z)}{4!} (-h^2) = -\frac{f''(z)h^2}{12} = O(h^2)$$

•  $f(z+h) = f(z) + f'(z)h + \frac{f''(\xi_1)}{2!} h^2$      f.s.  $\xi_1 \in (z, z+h)$  if  $f \in C^2([z, z+h])$

$\hookrightarrow f'(z) = \underbrace{\frac{f(z+h) - f(z)}{h}}_D - \underbrace{\frac{f''(\xi_1)}{2} h}_E$

•  $f(z-h) = f(z) - f'(z)h + \frac{f''(\xi_2)}{2!} h^2$      f.s.  $\xi_2 \dots$

$\hookrightarrow f'(z) = \underbrace{\frac{f(z) - f(z-h)}{h}}_D - \underbrace{\frac{f''(\xi_2)}{2} h}_E$

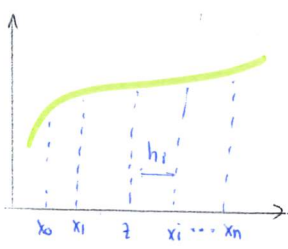
•  $f(z+h) = f(z) + f'(z)h + \frac{f''(\xi_1)}{2!} h^2 + \frac{f'''(\xi_1)}{3!} h^3$   
 •  $f(z-h) = f(z) - f'(z)h + \frac{f''(\xi_2)}{2!} h^2 - \frac{f'''(\xi_2)}{3!} h^3$

$f'(z) = \frac{f(z+h) - f(z-h)}{2h} - \frac{f''(\xi_1) + f''(\xi_2)}{2} \frac{h^2}{3!} = \underbrace{\frac{f(z+h) - f(z-h)}{2h}}_D - \underbrace{\frac{f'''(\xi)}{6} h^2}_E$

•  $f(z+h) = f(z) + \dots + \frac{f^{(3)}(z)}{3!} h^3 + \frac{f^{(4)}(\xi_1)}{4!} h^4$   
 •  $f(z-h) = f(z) - \dots - \frac{f^{(3)}(z)}{3!} h^3 + \frac{f^{(4)}(\xi_1)}{4!} h^4$

$f''(z) = \frac{f(z+h) - 2f(z) + f(z-h)}{h^2} - \frac{f^{(4)}(\xi)}{12} h^2$

TAYLOR GENERAL



•  $f(x_i) = f(z+h_i) = f(z) + f'(z)h_i + \frac{f''(z)}{2!} h_i^2 + \dots + \frac{f^{(n)}(z)}{n!} h_i^n + \dots + \frac{f^{(m)}(z)}{m!} h_i^m + \frac{f^{(m+1)}(\xi_i)}{(m+1)!} h_i^{m+1}$

$\begin{cases} x_i = z + h_i \\ f^{(k)}(z) = P_n^{(k)}(z) + E = \sum_{i=0}^n A_i f(x_i) + E \end{cases}$

$\Rightarrow f^{(k)}(z) = \sum_{i=0}^n A_i \cdot f(x_i) + E$

\* Assuming  $\begin{cases} n \geq k \\ m \geq n \end{cases} f \in C^{m+1}$

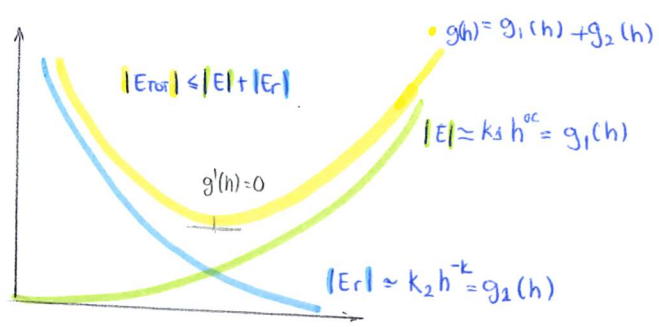
$$f^{(k)}(z) = f(z) \sum_{i=0}^n A_i + f'(z) \sum_{i=0}^n A_i h_i + f''(z) \frac{\sum_{i=0}^n A_i h_i^2}{2!} + \dots + f^{(k)}(z) \frac{\sum_{i=0}^n A_i h_i^k}{k!} + \dots + f^{(n)}(z) \frac{\sum_{i=0}^n A_i h_i^n}{n!} + f^{(n+1)}(z) \frac{\sum_{i=0}^n A_i h_i^{n+1}}{(n+1)!} + \dots$$

$$+ \dots + f^{(m)}(z) \frac{\sum_{i=0}^n A_i h_i^m}{(m)!} + f^{(m+1)}(z) \frac{\sum_{i=0}^n A_i h_i^{m+1}}{(m+1)!} + \epsilon$$

$$\begin{cases} \sum_{i=0}^n A_i = 0 \\ \sum_{i=0}^n A_i h_i = 0 \\ \sum_{i=0}^n A_i h_i^k = k! \\ \sum_{i=0}^n A_i h_i^n = 0 \end{cases} \rightarrow A_i$$

$$E = -\frac{f^{(n+1)}(z) \sum_{i=0}^n A_i h_i^{n+1}}{(n+1)!} - \dots - \frac{f^{(m)}(z) \sum_{i=0}^n A_i h_i^m}{m!} - \frac{1}{(m+1)!} \sum_{i=0}^n A_i f^{(m+1)}(\xi_i) h_i^{m+1}$$

OPTIMAL VALUE OF h (h<sub>opt</sub>)



$$D = \sum_{i=0}^n A_i f(x_i) \quad \left\{ \begin{array}{l} A_i = \frac{C_i}{h^k} \end{array} \right.$$

- $E \equiv$  truncation error
- $E_r \equiv$  rounding error
- $E_{tot} \equiv$  Total error ( $E_{tot} = E + E_r$ )

$$E_{tot} = f^{(k)}(z) - \bar{D} = f^{(k)}(z) - D + D - \bar{D}$$

not exact value

$$E_{tot} = f^{(k)}(z) - \bar{D} = \underbrace{f^{(k)}(z) - D}_E + \underbrace{D - \bar{D}}_{E_r}$$

Adding abs values:

$$|E_{tot}| = |f^{(k)}(z) - D + D - \bar{D}|$$

$$|E_r| = |D - \bar{D}| = \left| \sum_{i=0}^n A_i (f_i - \bar{f}_i) \right| \leq \underbrace{\sum_{i=0}^n A_i}_{\epsilon} |f_i - \bar{f}_i| = \epsilon \cdot A_f \rightarrow |E_r| \leq \sum A F = g_2(h)$$

UPPER BOUND OF ERROR IN NOODAL COORDINATES

Example

CALCULATE  $h_{opt}$  for:  $f'(z) = \frac{f(z+h) - f(z-h)}{2h} + \frac{(-h^2)}{6} f'''(\xi)$   
 $\hookrightarrow M \geq |f'''(\xi)|$

•  $|E_1| = \left| \frac{-h^2}{6} f'''(\xi) \right| \leq \frac{h^2}{6} M = g_1(h)$

•  $AF = \sum_{i=0}^{\infty} |A_i| = \left| \frac{1}{2h} \right| + \left| \frac{-1}{2h} \right| = \frac{1}{h}$

•  $|E_r| = \varepsilon \cdot AF = \frac{\varepsilon}{h} = g_2(h)$

•  $g(h) = g_1(h) + g_2(h) = \frac{h^2 M}{6} + \varepsilon h^{-1}$

$\hookrightarrow g'(h) = \frac{2hM}{6} - \varepsilon h^{-2} = 0 \rightarrow \frac{2hM}{3} = \frac{\varepsilon}{h^2} \rightarrow h^3 = \frac{3\varepsilon}{M} \rightarrow h = \sqrt[3]{\frac{3\varepsilon}{M}}$

$g(h_{opt}) = \frac{\sqrt[3]{9}}{2} \varepsilon^{2/3} \cdot M^{1/3}$

(IT HAS TO BE A SINGLE TERM)



EULER:  $y_{k+1} = y_k + f(t_k, y_k) h$

Ⓔ

ENHANCED EULER:  $y_{k+1} = y_k + \frac{(k_1 + k_2)}{2}$

$$\left\{ \begin{array}{l} k_1 = h \cdot f(t_k, y_k) \\ k_2 = h \cdot f(t_k + h, y_k + k_1) \end{array} \right.$$

Ⓕ

(MOD. EULER) MIDPOINT:  $y_{k+1} = y_k + k_2$

$$\left\{ \begin{array}{l} k_1 = h \cdot f(t_k, y_k) \\ k_2 = h \cdot f(t_k + h/2, y_k + k_1/2) \end{array} \right.$$

Ⓖ

RK4:  $y_{k+1} = y_k + \frac{k_1 + 2k_2 + 2k_3 + k_4}{6}$

$$\left\{ \begin{array}{l} k_1 = h \cdot f(t_k, y_k) \\ k_2 = h \cdot f(t_k + h/2, y_k + k_1/2) \\ k_3 = h \cdot f(t_k + h/2, y_k + k_2/2) \\ k_4 = h \cdot f(t_k + h, y_k + k_3) \end{array} \right.$$

Ⓖ

TRAPEZOIDAL:  $y_{k+1} = y_k + \frac{(f(t_k, y_k) + f(t_{k+1}, y_{k+1}))}{2} h$

Ⓙ

BACKWARD:  $y_{k+1} = y_k + f(t_{k+1}, y_{k+1}) h$  como EULER pero con  $f_{k+1}$

Ⓚ

$n$  PAR (EVEN)  $\Rightarrow \pi'(z) = 0$  SUFFICIENT

$z$  is a node  $\Leftrightarrow \pi(z) = 0$  NECESSARY AND SUFFICIENT

