

# ENGINEERING MATHEMATICS II

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**010.141**

**NUMERICAL METHODS  
IN  
GENERAL**

**MODULE 5**



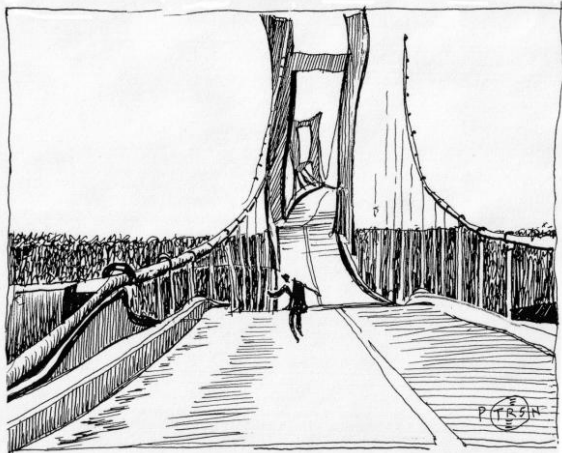
# NUMERICAL METHODS

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- Methods for solving problems numerically on a computer
- **Steps:**
  - Modeling,
  - Choice of a numerical method, Programming,
  - Doing the computation,
  - Interpreting the results
- Solution of equations by iteration
- Interpolation
- Numerical integration and differentiation



# Nonlinear Equations in Engineering Fields



$$f(x) = \omega_n(x) - \omega = 0$$

x: bridge design variables

To design a safe bridge, a natural frequency must be higher than that of an external loading.

$$f(x) = T(x) - T^* = 0$$

x: battery design variables

To design a safe battery, a temperature level must be smaller than a marginal temperature.



$$f(x) = \sigma(x) - S = 0$$

x: bridge gusset plate

To design a safe bridge, a stress level at a critical bridge element must be smaller than its strength.

# SOLUTION OF EQUATIONS BY ITERATION

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- Solving equation  $f(x) = 0$
- **Methods:**
  - Fixed – Point Iteration
  - Newton's Method
  - Secant Method



# FIXED-POINT ITERATION FOR SOLVING $f(x) = 0$

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- **Idea:** transform  $f(x) = 0$  into  $x = g(x)$
- **Steps:**
  1. Choose  $x_0$
  2. Compute  $x_1 = g(x_0)$ ,  $x_2 = g(x_1)$ , ...,  $x_{n+1} = g(x_n)$
- A solution of  $x = g(x)$  is called a **fixed point**
- Depending on the initial value chosen ( $x_0$ ), the related sequences may converge or diverge



# FIXED-POINT ITERATION FOR SOLVING $f(x) = 0$

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➤ **Example:**  $f(x) = x^2 - 3x + 1 = 0$

$$\text{Solutions} = \begin{cases} 2.618034 \\ 0.381966 \end{cases}$$

The equation may be written

$$(4a) \quad x = g_1(x) = \frac{1}{3}(x^2 + 1), \quad \text{thus} \quad x_{n+1} = \frac{1}{3}(x_n^2 + 1).$$

If we choose  $x_0 = 1$ , we obtain the sequence (Fig. 423a; computed with 6S and then rounded)

$$x_0 = 1.000, \quad x_1 = 0.667, \quad x_2 = 0.481, \quad x_3 = 0.411, \quad x_4 = 0.390, \dots$$

which seems to approach the smaller solution. If we choose  $x_0 = 2$ , the situation is similar. If we choose  $x_0 = 3$ , we obtain the sequence (Fig. 423a, upper part)

$$x_0 = 3.000, \quad x_1 = 3.333, \quad x_2 = 4.037, \quad x_3 = 5.766, \quad x_4 = 11.415, \dots$$

which diverges.



# FIXED-POINT ITERATION FOR SOLVING $f(x) = 0$

Our equation may also be written (divide by  $x$ )

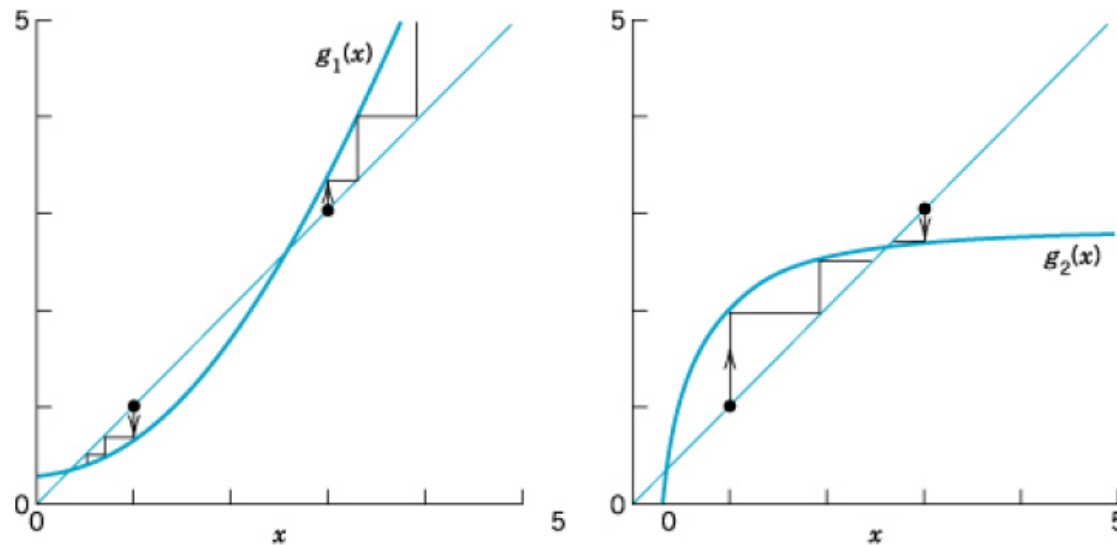
$$(4b) \quad x = g_2(x) = 3 - \frac{1}{x}, \quad \text{thus} \quad x_{n+1} = 3 - \frac{1}{x_n},$$

and if we choose  $x_0 = 1$ , we obtain the sequence (Fig. 423b)

$$x_0 = 1.000, \quad x_1 = 2.000, \quad x_2 = 2.500, \quad x_3 = 2.600, \quad x_4 = 2.615, \dots$$

which seems to approach the larger solution. Similarly, if we choose  $x_0 = 3$ , we obtain the sequence (Fig. 423b)

$$x_0 = 3.000, \quad x_1 = 2.667, \quad x_2 = 2.625, \quad x_3 = 2.619, \quad x_4 = 2.618, \dots$$



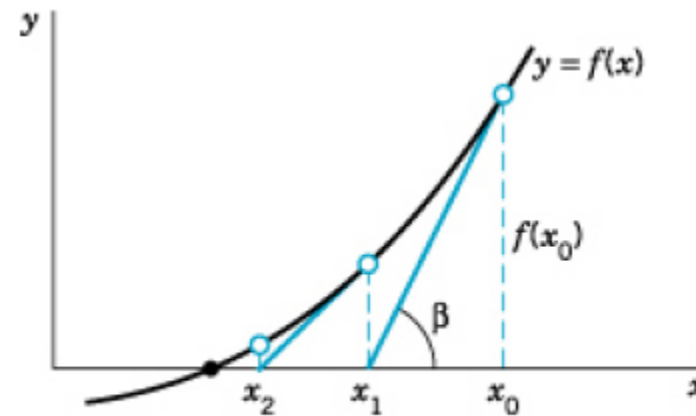
# NEWTON'S METHOD FOR SOLVING $f(x) = 0$

- $f$  must have a continuous derivative  $f'$
- The method is simple and fast

$$\tan \beta = f'(x_0) = \frac{f(x_0)}{x_0 - x_1}$$

⇓

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



**Fig. 425.** Newton's method



# NEWTON'S METHOD FOR SOLVING $f(x) = 0$

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

⋮

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



# NEWTON'S METHOD FOR SOLVING $f(x) = 0$

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**Example:** Find the positive solution of  $f(x) = 2 \sin x - x = 0$

**Solution:**

Setting  $f(x) = x - 2 \sin x$ , we have  $f'(x) = 1 - 2 \cos x$ , and (5) gives

$$x_{n+1} = x_n - \frac{x_n - 2 \sin x_n}{1 - 2 \cos x_n} = \frac{2(\sin x_n - x_n \cos x_n)}{1 - 2 \cos x_n} = \frac{N_n}{D_n}.$$

From the graph of  $f$  we conclude that the solution is near  $x_0 = 2$ . We compute:

$n$	$x_n$	$N_n$	$D_n$	$x_{n+1}$
0	2.00000	3.48318	1.83229	1.90100
1	1.90100	3.12470	1.64847	1.89552
2	1.89552	3.10500	1.63809	1.89550
3	1.89550	3.10493	1.63806	1.89549

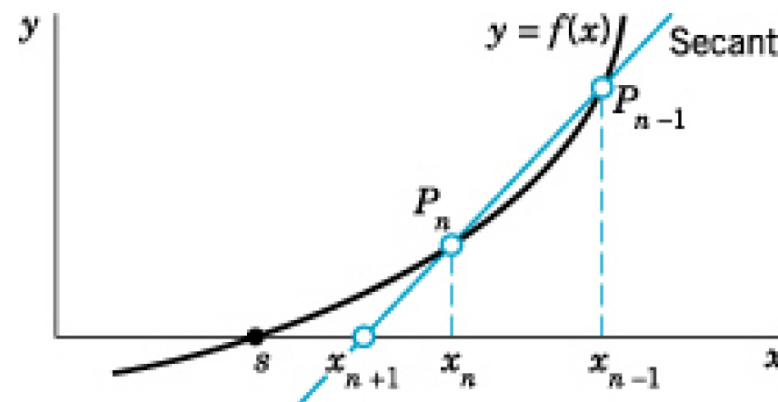
$x_4 = 1.89549$  is exact to 5D since the solution to 6D is 1.895 494.



# SECANT METHOD FOR SOLVING $f(x) = 0$

Newton's method is powerful but disadvantageous because it is difficult to obtain  $f'$ . The secant method approximates  $f'$ .

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



**Fig. 426.** Secant method

# SECANT METHOD FOR SOLVING $f(x) = 0$

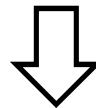
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➤ Newton's Method

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

➤ Secant

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



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



# SECANT METHOD FOR SOLVING $f(x) = 0$

Find the positive solution of  $f(x) = x - 2 \sin x = 0$  by the secant method, starting from  $x_0 = 2, x_1 = 1.9$ .

**Solution:**

Here, (10) is

$$x_{n+1} = x_n - \frac{(x_n - 2 \sin x_n)(x_n - x_{n-1})}{x_n - x_{n-1} + 2(\sin x_{n-1} - \sin x_n)} = x_n - \frac{N_n}{D_n}.$$

Numerical values are:

$n$	$x_{n-1}$	$x_n$	$N_n$	$D_n$	$x_{n+1} - x_n$
1	2.000 000	1.900 000	-0.000 740	-0.174 005	-0.004 253
2	1.900 000	1.895 747	-0.000 002	-0.006 986	-0.000 252
3	1.895 747	1.895 494	0		0

$x_3 = 1.895 494$  is exact to 6D. See Example 4.



진정성이 마음을 움직인다.

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송나라 범관, 계산행려도

# INTERPOLATION

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- Function  $f(x)$  is unknown
- Some values of  $f(x)$  are known  $(f_1, f_2, \dots, f_n)$
- **Idea:** Find a polynomial  $p_n(x)$  that is an approximation of  $f(x)$

$$p_n(x_0) = f_0, p_n(x_1) = f_1, \dots, p_n(x_n) = f_n$$

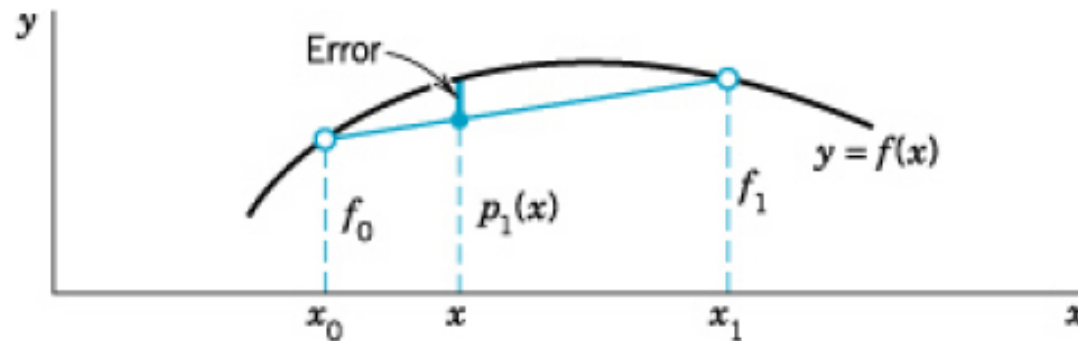
- **Lagrange interpolation**
  - Linear
  - Quadratic
  - General
- **Newton's interpolation**
  - Divided difference
  - Forward difference
  - Backward difference
- **Splines**



# LINEAR LAGRANGE INTERPOLATION

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- Use 2 known values of  $f(x) \rightarrow f_0, f_1$



**Fig. 428.** Linear interpolation

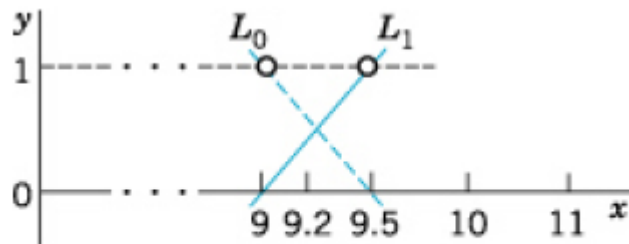
$p_1$  is the linear Lagrange polynomial.



# LINEAR LAGRANGE INTERPOLATION

- $p_1(x) = L_0(x)f_0 + L_1(x)f_1$
- $L_0$  and  $L_1$  are linear polynomials (weight functions).

$$L_0(x) = \begin{cases} 1 & \text{if } x = x_0 \\ 0 & \text{if } x = x_1 \end{cases} \quad L_1(x) = \begin{cases} 0 & \text{if } x = x_0 \\ 1 & \text{if } x = x_1 \end{cases}$$



**Fig. 429.**  $L_0$  and  $L_1$  in Example 1

$$L_0(x) = \frac{x - x_1}{x_0 - x_1}, \quad L_1(x) = \frac{x - x_0}{x_1 - x_0}$$

$$p_1(x) = \frac{x - x_1}{x_0 - x_1} f_0 + \frac{x - x_0}{x_1 - x_0} f_1$$

# LINEAR LAGRANGE INTERPOLATION

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Example: Compute  $\ln 9.2$  from  $\ln 9.0 = 2.1972$  and  $\ln 9.5 = 2.2513$  by linear Lagrange interpolation

Solution:

$$x_0 = 9.0, \quad x_1 = 9.5, \quad f_0 = \ln 9.0, \quad f_1 = \ln 9.5$$

$$L_0(9.2) = \frac{9.2 - 9.5}{9.0 - 9.5} = 0.6 \quad L_1(9.2) = \frac{9.2 - 9.0}{9.5 - 9.0} = 0.4$$

$$\ln 9.2 \approx p_1(9.2)$$

$$= L_0(9.2)f_0 + L_1(9.2)f_1$$

$$= 0.6(2.1972) + 0.4(2.2513)$$

$$= 2.2188$$

$$\text{Error: } \ln 9.2 - p_1(9.2) = 2.2192 - 2.2188 = 0.0004$$



# QUADRATIC LAGRANGE INTERPOLATION

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- Use of 3 known values of  $f(x) \rightarrow f_0, f_1, f_2$
- Approximation of  $f(x)$  by a second-degree polynomial

$$p_2(x) = L_0(x)f_0 + L_1(x)f_1 + L_2(x)f_2$$

$$L_0 = \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)}$$

$$L_1 = \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)}$$

$$L_2 = \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_2 - x_1)}$$



# QUADRATIC LAGRANGE INTERPOLATION

**Example:** Compute  $\ln 9.2$  for  $f_0(x_0 = 9.0) = \ln 9.0$ ,  $f_1(x_1 = 9.5) = \ln 9.5$ ,  $f_2(x_2 = 11.0) = \ln 11.0$

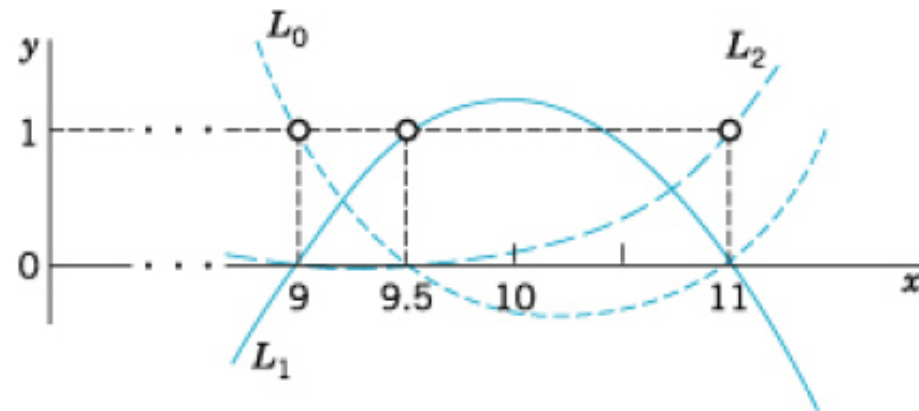
**Solution:**

$$L_0(x) = x^2 - 20.5x + 104.5$$

$$L_1(x) = -\frac{1}{0.75}(x^2 - 20x + 99)$$

$$L_2(x) = \frac{1}{3}(x^2 - 18.5x + 85.5)$$

$$\ln 9.2 \approx p_2(9.2) = 2.2192$$



**Fig. 430.**  $L_0, L_1, L_2$  in Example 2

# GENERAL LAGRANGE INTERPOLATION

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$$\begin{aligned} f(x) \approx p_n(x) &= \sum_{k=0}^n L_k(x) f_k \\ &= \sum_{k=0}^n \frac{l_k(x)}{l_k(x_k)} f_k \end{aligned}$$

$$l_k(x) = (x - x_0) \cdots (x - x_{k-1})(x - x_{k+1}) \cdots (x - x_n)$$



# NEWTON'S INTERPOLATION

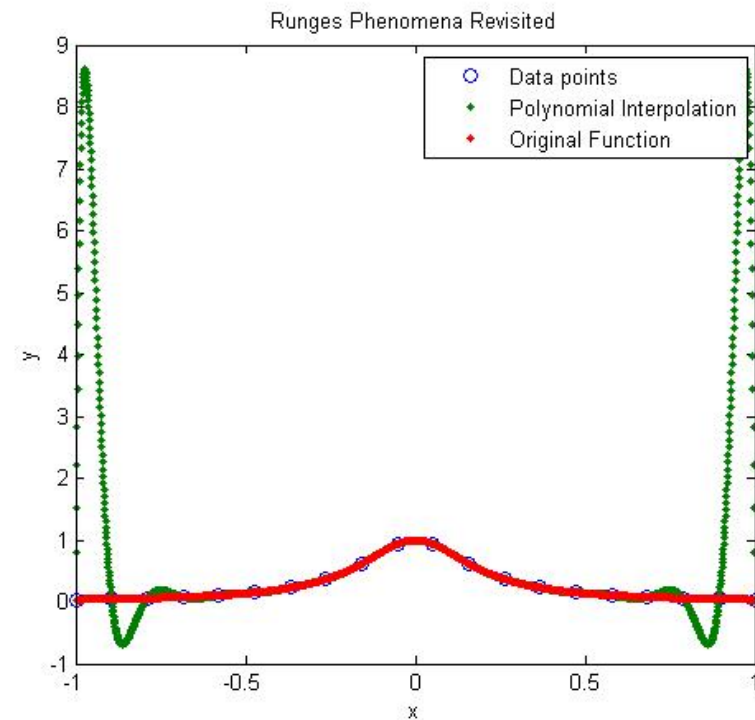
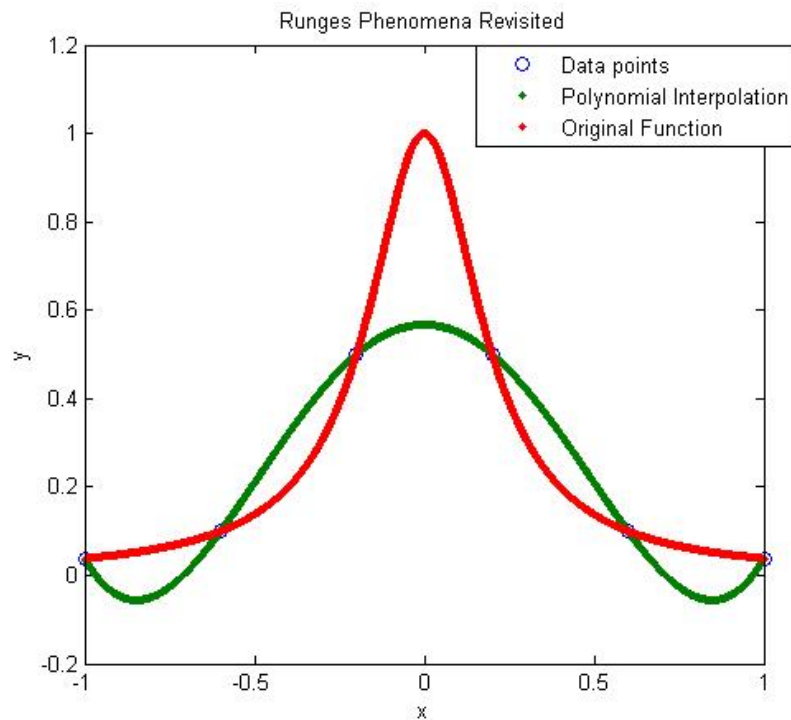
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- More appropriate for computation
- Level of accuracy can be easily improved by adding new terms that increase the degree of the polynomial
- Divided difference
- Forward difference
- Backward difference



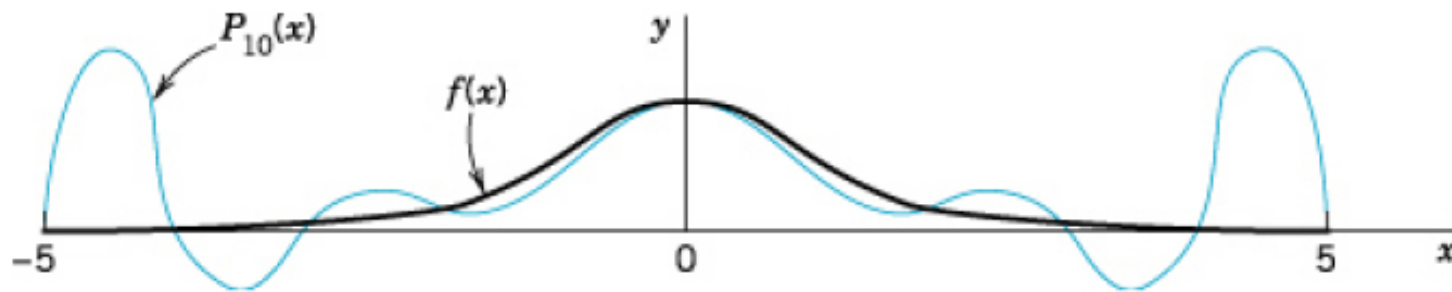
# Is High-Order Polynomial a Good Idea?

$$f(x) = 1/(1+25x^2)$$



# SPLINES

- Method of interpolation used to avoid numerical instability



**Fig. 431.** Runge's example  $f(x) = 1/(1 + x^2)$  and interpolating polynomial  $P_{10}(x)$

- **Idea:** given an interval  $[a, b]$  where the high-degree polynomial can oscillate considerable, we subdivide  $[a, b]$  in several smaller intervals and use several low-degree polynomials (which cannot oscillate much)



# SPLINES (cont)

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- In an interval  $x \in [x_j, x_{j+1}]$ ,  $j = 0, \dots, n-1$

$$p_j(x) = a_{j_0} + a_{j_1}(x - x_j) + a_{j_2}(x - x_j)^2 + a_{j_3}(x - x_j)^3$$

where

$$a_{j_0} = p_j(x_j) = f_j$$

$$a_{j_1} = p'_j(x_j) = k_j$$

$$a_{j_2} = \frac{1}{2} p''_j(x_j) = \frac{3}{h^2} (f_{j+1} - f_j) - \frac{1}{h} (k_{j+1} + 2k_j)$$

$$a_{j_3} = \frac{2}{h^3} (f_j - f_{j+1}) + \frac{1}{h^2} (k_{j+1} + k_j)$$



## SPLINES (cont)

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- Let  $(x_0, f_0), (x_1, f_1), \dots, (x_n, f_n)$ .
- $k_0, k_n$  are two given numbers.
- $k_1, k_2, \dots, k_{n-1}$  are determined by a linear system of  $n-1$  equations:

$$k_{j-1} + 4k_j + k_{j+1} = \frac{3}{h}(f_{j+1} - f_{j-1})$$
$$p'_j(x_j) = k_j, \quad p'_j(x_{j+1}) = k_{j+1} \quad (j = 0, 1, \dots, n-1)$$

- $h$  is the distance between nodes

$$x_n = x_0 + nh$$



# SPLINES (cont)

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- Clamped conditions

$$g'(x_0) = f'(x_0), \quad g'(x_n) = f'(x_n)$$

- Free/natural conditions

$$g''(x_0) = 0, \quad g''(x_n) = 0$$

- **Example:**

Interpolate  $f(x) = x^4$  on interval  $x \in [-1, 1]$  by cubic spline in partitions  $x_0 = -1, x_1 = 0, x_2 = 1$ , satisfying clamped conditions

$$g'(-1) = f'(-1), \quad g'(1) = f'(1)$$



# SPLINES (cont)

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the given data are  $f_0 = f(-1) = 1, f_1 = f(0) = 0, f_2 = f(1) = 1$ .

$$q_0(x) = a_{00} + a_{01}(x + 1) + a_{02}(x + 1)^2 + a_{03}(x + 1)^3 \quad (-1 \leq x \leq 0)$$
$$q_1(x) = a_{10} + a_{11}x + a_{12}x^2 + a_{13}x^3 \quad (0 \leq x \leq 1)$$

$$k_0 + 4k_1 + k_2 = 3(f_2 - f_0).$$

Here  $f_0 = f_2 = 1$  (the value of  $x^4$  at the ends) and  $k_0 = -4, k_2 = 4$ .

$$-4 + 4k_1 + 4 = 3(1 - 1) = 0, \quad k_1 = 0$$

$$a_{00} = f_0 = 1, a_{01} = k_0 = -4$$

$$a_{02} = \frac{3}{1^2}(f_1 - f_0) - \frac{1}{1}(k_1 + 2k_0) = 3(0 - 1) - (0 - 8) = 5$$

$$a_{03} = \frac{2}{1^3}(f_0 - f_1) + \frac{1}{1^2}(k_1 + k_0) = 2(1 - 0) + (0 - 4) = -2.$$



# SPLINES (cont)

Similarly, for the coefficients of  $q_1$  we obtain from (13) the values  $a_{10} = f_1 = 0$ ,  $a_{11} = k_1 = 0$ , and

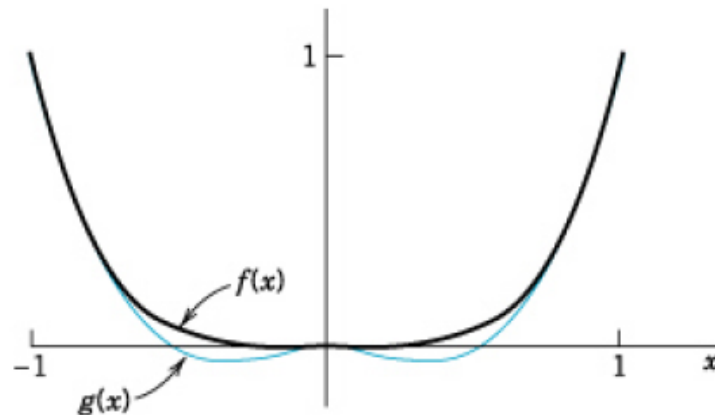
$$a_{12} = 3(f_2 - f_1) - (k_2 + 2k_1) = 3(1 - 0) - (4 + 0) = -1$$

$$a_{13} = 2(f_1 - f_2) + (k_2 + k_1) = 2(0 - 1) + (4 + 0) = 2.$$

This gives the polynomials of which the spline  $g(x)$  consists, namely,

$$g(x) = \begin{cases} q_0(x) = 1 - 4(x + 1) + 5(x + 1)^2 - 2(x + 1)^3 = -x^2 - 2x^3 & \text{if } -1 \leq x \leq 0 \\ q_1(x) = -x^2 + 2x^3 & \text{if } 0 \leq x \leq 1. \end{cases}$$

Figure 433 shows  $f(x)$  and this spline. Do you see that we could have saved over half of our work by using symmetry?

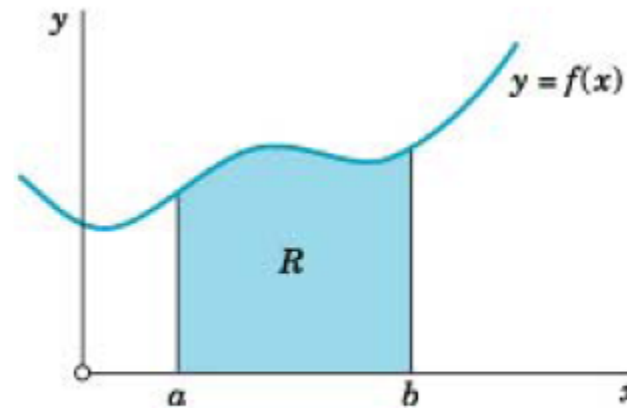


**Fig. 433.** Function  $f(x) = x^4$  and cubic spline  $g(x)$  in Example 1

# NUMERICAL INTEGRATION

- Numerical evaluation of integrals whose analytical evaluation is too complicated or impossible, or that are given by recorded numerical values

$$\int_a^b f(x) dx$$



**Fig. 437.** Geometric interpretation of a definite integral

- Rectangular rule
- Trapezoidal rule
- Simpson's rule
- Gauss integration



# RECTANGULAR RULE

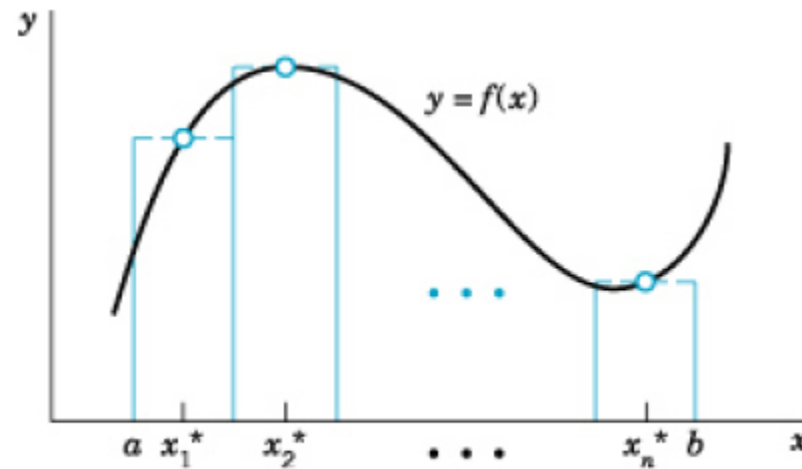
- Approximation by n rectangular areas

$$J = \int_a^b f(x) dx \approx h [f(x^*_1) + f(x^*_2) + \dots + f(x^*_n)]$$

where

$$h = \frac{b-a}{n}$$

$$x_1 = x_0 + h$$

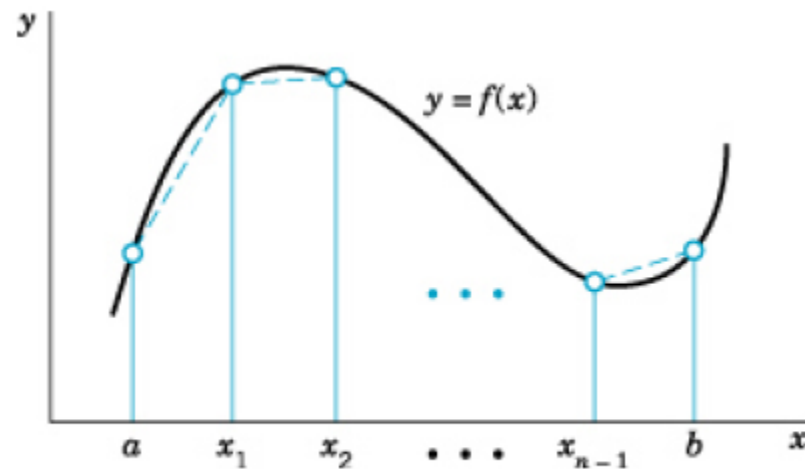


**Fig. 438.** Rectangular rule

# TRAPEZOIDAL RULE

- Approximation by n trapezoidal areas

$$J = \int_a^b f(x) dx \approx h \left[ \frac{1}{2} f(a) + f(x_1) + f(x_2) + \cdots + f(x_{n-1}) + \frac{1}{2} f(b) \right]$$



**Fig. 439.** Trapezoidal rule



# TRAPEZOIDAL RULE

Evaluate  $J = \int_0^1 e^{-x^2} dx$  by means of (2) with  $n = 10$ .

**Solution:**

$J \approx 0.1(0.5 \cdot 1.367\ 879 + 6.778\ 167) = 0.746\ 211$  from Table 19.3.

**TABLE 19.3** Computations in Example 1

$j$	$x_j$	$x_j^2$	$e^{-x_j^2}$	$j$	$x_j$	$x_j^2$	$e^{-x_j^2}$
0	0	0	1.000 000	6	0.6	0.36	0.697 676
1	0.1	0.01	0.990 050	7	0.7	0.49	0.612 626
2	0.2	0.04	0.960 789	8	0.8	0.64	0.527 292
3	0.3	0.09	0.913 931	9	0.9	0.81	0.444 858
4	0.4	0.16	0.852 144	10	1.0	1.00	0.367 879
5	0.5	0.25	0.778 801	Sums		1.367 879	6.778 167



# SIMPSON'S RULE

- Approximation by parabolas using Lagrange polynomials  $p_2(x)$
- Interval of integration  $[a, b]$  divided into an even number of subintervals

$$h = \frac{b-a}{2m} \quad f_0 = f(x_0) \quad x_1 = a + h \quad x_2 = x_1 + h$$

$$\int_a^b f(x) dx \approx \frac{h}{3} (f_0 + 4f_1 + 2f_2 + 4f_3 + \cdots + 2f_{2m-2} + 4f_{2m-1} + f_{2m})$$

$$f_j = f(x_j)$$

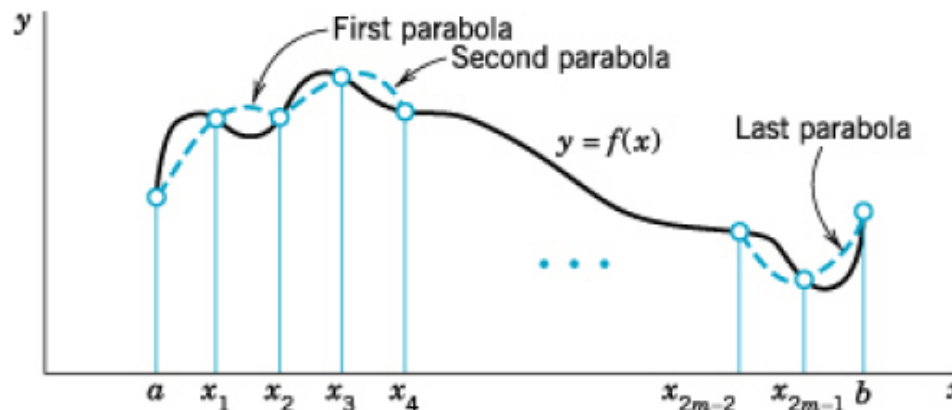


Fig. 440. Simpson's rule

# SIMPSON'S RULE

Evaluate  $J = \int_0^1 e^{-x^2} dx$  by Simpson's rule with  $2m = 10$  and estimate the error.

**Solution:**

Since  $h = 0.1$ , Table 19.5 gives

$$J \approx \frac{0.1}{3}(1.367\ 879 + 4 \cdot 3.740\ 266 + 2 \cdot 3.037\ 901) = 0.746\ 825.$$

**TABLE 19.5** Computations in Example 3

$j$	$x_j$	$x_j^2$	$e^{-x_j^2}$	$j$	$x_j$	$x_j^2$	$e^{-x_j^2}$
0	0	0	1.000 000	6	0.6	0.36	0.697 676
1	0.1	0.01	0.990 050	7	0.7	0.49	0.612 626
2	0.2	0.04	0.960 789	8	0.8	0.64	0.527 292
3	0.3	0.09	0.913 931	9	0.9	0.81	0.444 858
4	0.4	0.16	0.852 144	10	1.0	1.00	0.367 879
5	0.5	0.25	0.778 801				
				Sums			1.367 879 3.740 266 3.037 901



# SIMPSON'S RULE

ALGORITHM SIMPSON ( $a, b, m, f_0, f_1, \dots, f_{2m}$ )

This algorithm computes the integral  $J = \int_a^b f(x) dx$  from given values  $f_j = f(x_j)$  at equidistant  $x_0 = a, x_1 = x_0 + h, \dots, x_{2m} = x_0 + 2mh = b$  by Simpson's rule (7), where  $h = (b - a)/(2m)$ .

INPUT:  $a, b, m, f_0, \dots, f_{2m}$

OUTPUT: Approximate value  $\tilde{J}$  of  $J$

Compute  $s_0 = f_0 + f_{2m}$

$$s_1 = f_1 + f_3 + \dots + f_{2m-1}$$

$$s_2 = f_2 + f_4 + \dots + f_{2m-2}$$

$$h = (b - a)/2m$$

$$\tilde{J} = \frac{h}{3} (s_0 + 4s_1 + 2s_2)$$

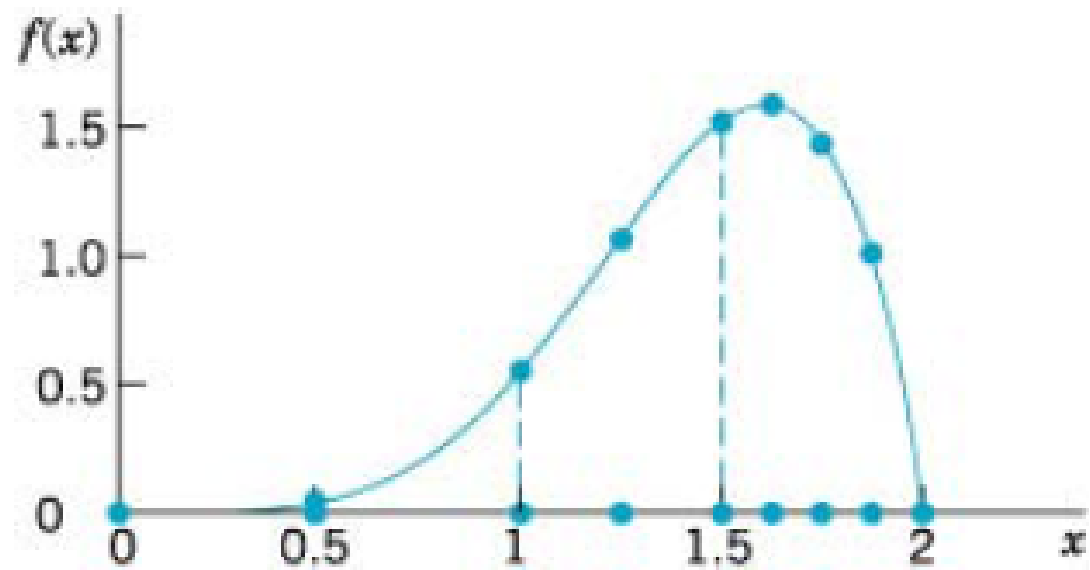
OUTPUT  $\tilde{J}$ . Stop.

End SIMPSON



# ADAPTIVE INTEGRATION

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**Fig. 441.** Adaptive integration in Example 6

# GAUSS INTEGRATION

$$\int_a^b f(x) dx = \int_{-1}^1 f(t) dt \approx \sum_{j=1}^{\infty} A_j f_j$$

where

$$x = \frac{1}{2}[a(t-1) + b(t+1)]$$

$A_1, \dots, A_n \Rightarrow$  coefficients

$$f_j = f(t_j)$$

**TABLE 19.7** Gauss Integration: Nodes  $t_j$  and Coefficients  $A_j$

$n$	Nodes $t_j$	Coefficients $A_j$	Degree of Precision
2	-0.57735 02692	1	3
	0.57735 02692	1	
3	-0.77459 66692	0.55555 55556	5
	0	0.88888 88889	
	0.77459 66692	0.55555 55556	
4	-0.86113 63116	0.34785 48451	7
	-0.33998 10436	0.65214 51549	
	0.33998 10436	0.65214 51549	
	0.86113 63116	0.34785 48451	
5	-0.90617 98459	0.23692 68851	9
	-0.53846 93101	0.47862 86705	
	0	0.56888 88889	
	0.53846 93101	0.47862 86705	
	0.90617 98459	0.23692 68851	

