

Similarly, putting $x = x_s$, $y = f_s$ in eqn ③, we get,

$$q_1 = \frac{f_1}{(x_1 - x_0)(x_1 - x_2) \dots (x_1 - x_n)}$$

Substituting the value in ②, we get eqn ①.

Note: With $(n+1)$ sample points, we can generate n^{th} degree polynomial.

The equation ① may be represented in general form as

$$f(x) = \sum_{i=0}^n f_i l_i(x) \quad \dots \quad (3)$$

$$\text{where, } l_i(x) = \frac{\prod_{j=0, j \neq i}^{n-1} (x - x_j^o)}{(x_i^o - x_j^o)}$$

Equation (3) is called Lagrange interpolation polynomials.

Q. The table below gives the square root of integers.

x	1	2	3	4	5
$f(x)$	1	1.4142	1.7321	2	2.2361

find the square root of 2.5 using second order Lagrange interpolation polynomial.

Soln

Let us consider the following three points
 $x_1 = 2$, $x_2 = 3$ and $x_3 = 4$

$$x_0 = 2, \quad x_1 = 3 \quad \text{and} \quad x_2 = 4$$

$$\text{So, } f_0 = 1.4142 \quad f_1 = 1.7321 \quad \text{and} \quad f_2 = 2$$

and $x = 2.5$

Then,

for $x = 2.5$, we have

$$\begin{aligned}
 l_0(2.5) &= \frac{(x - x_0)(x - x_1)}{(x_0 - x_1)(x_0 - x_2)} \\
 &= \frac{(2.5 - 3)(2.5 - 4)}{(2 - 3)(2 - 4)} \\
 &= \frac{(-0.5)(-1.5)}{(-1)(-2)} \\
 &= 0.375
 \end{aligned}$$

$$\begin{aligned}
 \text{Again, } l_1(2.5) &= \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)} \\
 &= \frac{(2.5 - 2)(2.5 - 4)}{(3 - 2)(3 - 4)} \\
 &= \frac{(0.5)(-1.5)}{(-1)(-2)} \\
 &= 0.75
 \end{aligned}$$

$$\begin{aligned}
 \text{Again } l_2(2.5) &= \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_2 - x_1)} \\
 &= \frac{(2.5 - 2)(2.5 - 3)}{(4 - 2)(4 - 3)} \\
 &= \frac{(0.5)(-0.5)}{2 \times 1} \\
 &= -0.125
 \end{aligned}$$

Then, the second order lagrange interpolation is given by
we have,

$$\begin{aligned}
 P_2(2.5) &= f_0 l_0 + f_1 l_1 + f_2 l_2 \\
 &= 1.4142 \times 0.375 + 1.7321 \times 0.75 + (-0.125) \times 2 \\
 &= 0.530325 + 1.299075 - 0.25 \\
 &= 1.579425 \\
 &\approx 1.5794
 \end{aligned}$$

Q. Find the Lagrange interpolation polynomial to fit the following data :

?	0	1	2	3
x_i	0	1	2	3
e^{x_i-1}	0	1.4183	6.3891	19.0855

Use the polynomial to estimate the value of $x \cdot e^{10.5}$

Here, $x_0 = 0$, $x_1 = 1$, $x_2 = 2$ and $x_3 = 3$

Now,

$$\begin{aligned} l_0(x) &= \frac{(x-x_1)(x-x_2)(x-x_3)}{(x_0-x_1)(x_0-x_2)(x_0-x_3)} \\ &= \frac{(x-1)(x-2)(x-3)}{(0-1)(0-2)(0-3)} \\ &= \frac{(x-1)(x-2)(x-3)}{-6} = \frac{1}{6}(x^3 - 6x^2 + 11x - 6) \end{aligned}$$

$$\begin{aligned} l_1(x) &= \frac{(x-x_0)(x-x_2)(x-x_3)}{(x_1-x_0)(x_1-x_2)(x_1-x_3)} \\ &= \frac{(x-0)(x-2)(x-3)}{(1-0)(1-2)(1-3)} \\ &= \frac{(x-0)(x-2)(x-3)}{2} = \frac{1}{2}(x^3 - 3x^2 + 2x) \end{aligned}$$

$$\begin{aligned} l_2(x) &= \frac{(x-x_0)(x-x_1)(x-x_3)}{(x_2-x_0)(x_2-x_1)(x_2-x_3)} \\ &= \frac{(x-0)(x-1)(x-3)}{(2-0)(2-1)(2-3)} \\ &= \frac{x(x-1)(x-3)}{-2} = \frac{1}{2}(x^3 - 4x^2 + 3x) \end{aligned}$$

$$\begin{aligned}
 L_3(x) &= \frac{(x-x_0)(x-x_1)(x-x_2)}{(x_3-x_0)(x_3-x_1)(x_3-x_2)} \\
 &= \frac{(x-0)(x-1)(x-2)}{(3-0)(3-1)(3-2)} \\
 &= \frac{x(x-1)(x-2)}{6}
 \end{aligned}$$

The, interpolating polynomial is,

$$\begin{aligned}
 p(x) &= f_0 L_0(x) + f_1 L_1(x) + f_2 L_2(x) + f_3 L_3(x) \\
 &= 1.7183 \times \frac{x(x-1)(x-2)}{2} + 6.3891 \times \frac{x(x-1)(x-3)}{-2} + \\
 &\quad 19.0855 \times \frac{x(x-1)(x-2)}{6} \\
 &= 1.7183 \times \frac{(x^2-2x)(x-3)}{2} + \frac{6.3891}{-2} \times (x^2-1)(x-3) + \\
 &\quad \frac{19.0855}{6} \times (x^2-1)(x-2) \\
 &= \frac{1.7183}{2} \times (x^3 - 2x^2 - 3x^2 + 6x) + \frac{6.3891}{-2} \times (x^3 - x - 3x + 3) + \\
 &\quad \frac{19.0855}{6} \times (x^3 - x - 2x^2 + 2)
 \end{aligned}$$

Newton's Interpolation Polynomial:

Divide Difference.

Let $(x_0, f_0), (x_1, f_1), (x_2, f_2), \dots, (x_n, f_n)$ be the given points then divided difference can be defined as follows.

\Rightarrow First divided difference for x_0, x_1 is given by

$$f[x_0, x_1] = \frac{f_1 - f_0}{x_1 - x_0}$$

\Rightarrow Second divided difference is given by.

$$\begin{aligned} f[x_0, x_1, x_2] &= \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0} \\ &= \frac{\frac{f_2 - f_1}{x_2 - x_1} - \frac{f_1 - f_0}{x_1 - x_0}}{x_2 - x_0} \end{aligned}$$

\Rightarrow Third divided difference is given by

$$\begin{aligned} f[x_0, x_1, x_2, x_3] &= \frac{f[x_2, x_3] - f[x_0, x_1, x_2]}{x_3 - x_0} \\ &= \frac{f[x_2, x_3] - f[x_1, x_2] - f[x_0, x_1] - f[x_0, x_2]}{x_3 - x_0} \end{aligned}$$

The newton's polynomial is,

$$P_n(x) = q_0 + q_1(x - x_0) + q_2(x - x_0)(x - x_1) + \dots + q_n(x - x_0)(x - x_1)\dots(x - x_{n-1}) \quad (1)$$

where the interpolation points x_0, x_1, \dots, x_{n-1} acts as centres.

Proof :-

To construct the interpolation polynomial, we need to determine the coefficient q_0, q_1, \dots, q_n .

Let us assume that, $(x_0, f_0), (x_1, f_1), \dots, (x_{n-1}, f_{n-1})$ are the interpolating points.

That is,

$$P_n(x_k) = f_k, \quad k = 0, 1, \dots, n-1 \quad \text{at } x = x_k$$

Now,

at $x = x_0$ we have from eqn ①

$$P_n(x_0) = f_0 = g_0 \quad \dots \quad ②$$

$$P_n(x_1) = f_1.$$

Similarly at $x = x_1$ we have,

$$P_n(x_1) = g_0 + g_1(x_1 - x_0) = f_1$$

substituting for g_0 from eqn ② we get,

$$g_1 = \frac{f_1 - f_0}{x_1 - x_0} \quad \dots \quad ③$$

at $x = x_2$,

$$P_n(x_2) = g_0 + g_1(x_2 - x_0) + g_2(x_2 - x_1)(x_2 - x_0) = f_2$$

substituting for g_0, g_1 and solving we get

$$g_2 = \frac{f_2 - f_1}{x_2 - x_1} - \frac{f_1 - f_0}{x_2 - x_0} \quad \dots \quad ④$$

let us define a notation,

$$f[x_k] = f_k$$

$$f[x_k, x_{k+1}] = \frac{f(x_{k+1}) - f(x_k)}{x_{k+1} - x_k}$$

Similarly,

$$f[x_k, x_{k+1}, x_{k+2}] = \frac{f(x_{k+2}) - f(x_k)}{x_{k+2} - x_k}$$

$$f[x_k, x_{k+1}, \dots, x_l, x_{l+1}] = \frac{f(x_{l+1}) - f(x_k)}{x_{l+1} - x_k} \quad \dots \quad ⑤$$

These quantities are called divided differences. Now we can express the coefficients a_i in terms of divided differences,

$$a_0 = f_0 = f[x_0]$$

$$a_1 = \frac{f_1 - f_0}{x_1 - x_0} = f[x_0, x_1]$$

$$a_2 = \frac{f_2 - f_1}{x_2 - x_1} - \frac{f_1 - f_0}{x_1 - x_0} = f[x_0, x_1, x_2]$$

Thus,

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

Here a_1 represent the first divided difference, a_2 represent the second divided difference and so on, substituting for a_i coefficient in eqⁿ ① we get,

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

which can be written as,

$$P_n(x) = \sum_{i=1}^n f[x_0, \dots, x_i] \prod_{j=0}^{i-1} (x - x_j) \quad \text{--- (6)}$$

which is the required Newton's Interpolation polynomial.

Q Given below table for $\log x$. Estimate $\log 2.5$ using second order Newton's Interpolation polynomial.

x_i	0	1	2	3
$\log x_i$	0	0.3010	0.4771	0.6021
	f_0	f_1	f_2	

Construction of divided difference table:

As we know that Newton's interpolation polynomial are evaluated using the divided difference at the interpolating points, we also have used the lower order divided difference to construct the higher order divided difference.

For example, consider the second order divided difference,

$$g_2 = f(x_0, x_1, x_2)$$

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

where $f[x_1, x_2]$ and $f[x_0, x_1]$ are first order divided difference, and are given by,

$$f[x_0, x_1] = \frac{f_1 - f_0}{x_1 - x_0}$$

$$\text{and, } f[x_1, x_2] = \frac{f_2 - f_1}{x_2 - x_1}$$

This shows that, given the interpolating points, we can obtain recursively a higher order divided difference starting from the first order difference.

for example, a divided difference table for 6 points is given below.

i	x_i	$f[x_i]$	first difference	Second difference	Third difference	Fourth difference	fifth difference
0	$f[x_0]$	$f[x_0]$	$f[x_0, x_1]$				
1	$f[x_1]$	$f[x_1]$		$f[x_0, x_1, x_2]$			
2	$f[x_2]$	$f[x_2]$	$f[x_1, x_2]$		$f[x_0, x_1, x_2, x_3]$		
3	$f[x_3]$	$f[x_3]$	$f[x_2, x_3]$	$f[x_1, x_2, x_3]$		$f[x_0, x_1, x_2, x_3, x_4]$	
4	$f[x_4]$	$f[x_4]$	$f[x_3, x_4]$	$f[x_2, x_3, x_4]$	$f[x_1, x_2, x_3, x_4]$		

Q Given the following set of data points, obtain the table of divided difference. Use the table to estimate the value of $f(1.5)$

i	0	1	2	3	4
x_i	1	2	3	4	5
$f(x_i)$	0	7	26	63	124

Sol,

The required table of divided difference is given below.

i	x_i	$f(x_i)$	first difference	Second difference	Third difference
0	1	0	7		
1	2	7	19	6	
2	3	26	37	9	1
3	4	63	61	12	
4	5	124			

Now, we know,

$$P_n(x) = q_0 + q_1 [x - x_0] + q_2 [x - x_0][x - x_1] + \dots$$

from the table, we have,

$$q_0 = 0$$

$$q_1 = 7$$

$$q_2 = 6$$

$$q_3 = 1$$

Now, from the formula

$$\begin{aligned}
 P_3(1.5) &= q_0 + q_1 [x - x_0] + q_2 [x - x_0][x - x_1] + q_3 [x - x_0][x - x_1][x - x_2] \\
 &= 0 + 7[1.5 - 1] + 6[1.5 - 1][1.5 - 2] + 1[1.5 - 1][1.5 - 2][1.5 - 3] \\
 &= 2.375 //
 \end{aligned}$$

Q4 Using the following table find $f(x)$ as a polynomial in x .

n	$f(n)$
-1	3
0	-6
3	39
6	822
7	1611

Sol:

Constructing the divided difference table, we have,

i	n_i	$f(n_i)$	first difference	Second difference	Third difference	fourth difference
0	-1	3				
1	0	-6	-9			
2	3	39	15	6		
3	6	822	261	41	5	
4	7	1611	789	13		

Now, we know,

$$P_0(x) = q_0 + q_1(x-n_0) + q_2(x-n_0)(x-n_1) + \dots$$

and from the table, we know

$$q_0 = 3$$

$$q_1 = -9$$

$$q_2 = 6$$

$$q_3 = 5$$

$$q_4 = 1$$

Then, we have,

$$\begin{aligned} P_0(x) &= q_0 + q_1(x-n_0) + q_2(x-n_0)(x-n_1) + q_3(x-n_0)(x-n_1)(x-n_2) \\ &\quad + q_4(x-n_0)(x-n_1)(x-n_2)(x-n_3) \\ &= 3 + -9(x+1) + q_2(x+1)(x-0) + q_3(x+1)(x-0)(x-3) \\ &\quad + q_4(x+1)(x-0)(x-3)(x-6) \end{aligned}$$

$$\begin{aligned}
 &= f + a_1x(x+1) + a_2x(x+1)(x-3) + a_3x(x+1)(x-3)(x-6) \\
 &= 3 + 9x + 9 + 6x(x+1) + 5x(x^2 - 3x + x - 3) + x(x^2 - 3x + x - 3) \\
 &= -6x - 9x + 6x^2 + 6x + 5x^3 - 15x^2 + 5x^2 - 15x + x(x^3 - 3x^2 \\
 &\quad + x^2 - 3x - 6x^2 + 18x + 6x + 18) \\
 &= -6x - 9x + 6x^2 + 6x + 5x^3 - 15x^2 + 5x^2 - 15x + x^4 - 3x^3 \\
 &\quad + x^3 - 3x^2 - 6x^3 + 18x^2 - 6x^2 + 18x \\
 &= x^4 - 3x^3 + 5x^2 + 6x
 \end{aligned}$$

Interpolation with equidistant points:

- Equally spaced data can be represented by,
 $x_{i+1} = x_i + h$ or $x_k = x_0 + kh$
 where h is the step size and k is any integer (+ve or -ve)
- Also forward difference can be defined as,

$$\Delta^j f_i = \Delta^{j-1} f_{i+1} - \Delta^{j-1} f_i$$

first forward difference

$$\Delta^2 f_i = \Delta f_{i+1} - \Delta f_i$$

second forward difference

$$\Delta^3 f_i = \Delta f_{i+1} - \Delta f_i$$

In tabular form, we have

x_i	f_i	1 st diff (Δf_i)	Second diff ($\Delta^2 f_i$)	3 rd diff ($\Delta^3 f_i$)	4 th diff ($\Delta^4 f_i$)
x_0	f_0				
x_1	f_1	$\Delta f_0 = f_1 - f_0$	$\Delta^2 f_0 = \Delta f_1 - \Delta f_0$		
x_2	f_2	$\Delta f_1 = f_2 - f_1$	$\Delta^2 f_1 = \Delta f_2 - \Delta f_1$	$\Delta^3 f_0 = \Delta^2 f_1 - \Delta^2 f_0$	$\Delta^4 f_0 = \Delta^3 f_1 - \Delta^3 f_0$
x_3	f_3	$\Delta f_2 = f_3 - f_2$	$\Delta^2 f_2 = \Delta f_3 - \Delta f_2$	$\Delta^3 f_1 = \Delta^2 f_2 - \Delta^2 f_1$	
x_4	f_4	$\Delta f_3 = f_4 - f_3$			

Here,

$$H_1 = H_0 + h$$

$$H_2 = H_1 + h \text{ and so on..}$$

We have,

$$P_n(x) = f_0 + q_1(x-H_0) + q_2(x-H_1)(x-H_0) + \dots$$

where,

$$q_0 = f_0$$

$$q_1 = F(H_0, H_1) = \frac{f_1 - f_0}{H_1 - H_0} = \frac{\Delta f_0}{h}$$

$$q_2 = F(H_0, H_1, H_2) = \frac{F(H_1, H_2) - F(H_0, H_1)}{H_2 - H_0} =$$

$$= \frac{f_2 - f_1}{H_2 - H_1} - \frac{f_1 - f_0}{H_2 - H_0}$$

$$= \frac{\Delta f_1 - \Delta f_0}{\Delta h}$$

$$= \frac{\Delta^2 f_0}{\Delta h^2}$$

$$\therefore \Delta f_1 - \Delta f_0 = \Delta^2 f_0$$

$$= \frac{\Delta^2 f_0}{2h^2}$$

Similarly,

$$q_3 = \frac{\Delta^3 f_0}{3! h^3}$$

So,

The general term is,

$$q_n = \frac{\Delta^n f_0}{n! h^n}$$

Substituting these value in Newton's divided difference interpolation polynomial, we get.

$$P_n(x) = f_0 + \frac{\Delta f_0}{h} (x-H_0) + \frac{\Delta^2 f_0}{2h^2} (x-H_0)(x-H_1) + \dots$$

This equation is known as Gregory-Newton's forward difference formula (Evenly spaced data)

Q) Estimate the value of $\sin \theta$ at $\theta = 25^\circ$ using Newton-Gregory forward difference formula with the help of following table.

0	10	20	30	40	50
$\sin \theta$	0.1736	0.3420	0.5000	0.6428	0.7660

Soln,

In order to use Newton-Gregory forward difference formula, we need the value of, $\Delta^4 f_0$. These coefficients can be obtained from the difference table as follows.

$x(\theta)$	$f(x)\sin \theta$	1st diff Δf_0	2nd diff $\Delta^2 f_0$	3rd diff $\Delta^3 f_0$	4th diff $\Delta^4 f_0$
10	0.1736	0.1684			
20	0.3420	-0.0104		-0.0048	
30	0.5000	-0.0152		-0.0004	
40	0.6428	-0.0196	-0.0044		
50	0.7660	0.0132			

Then we have;

$$\begin{aligned}
 P_n(x) &= f_0 + \frac{\Delta f_0}{h} (n-n_0) + \frac{\Delta^2 f_0}{2! h^2} (n-n_0)(n-n_1) \\
 &\quad + \frac{\Delta^3 f_0}{3! h^3} (n-n_0)(n-n_1)(n-n_2) + \frac{\Delta^4 f_0}{4! h^4} (n-n_0)(n-n_1)(n-n_2)(n-n_3) \\
 &= 0.1736 + \frac{0.1684}{10} (25-10) + \frac{(-0.0104)}{2 \times 10^2} (25-10)(25-20) \\
 &\quad + \frac{(-0.0048)}{3! 10^3} (25-10)(25-20)(25-30) + \frac{0.0004}{4! \times 10^4} (25-10)(25-20) \\
 &\quad \cdot (25-30) (25-40)
 \end{aligned}$$

$$\begin{aligned}
 &= 0.1786 + 0.2526 - 0.0039 + 0.003 + 0.0009375 \\
 &= 0.4256375.
 \end{aligned}$$

Backward difference.

If the table is too long and if the required point is closed to the end of the table, we can use another formula known as Newton-Gregory backward difference formula. The table for Newton-Gregory backward difference is as follows.

x_i	f	Δf	$\Delta^2 f$	$\Delta^3 f$	$\Delta^4 f$
x_0	f_0	Δf_1			
x_1	f_1	Δf_2	$\Delta^2 f_2$	$\Delta^3 f_3$	
x_2	f_2	Δf_3	$\Delta^2 f_3$	$\Delta^3 f_4$	$\Delta^4 f_4$
x_3	f_3	Δf_4	$\Delta^2 f_4$		
x_4	f_4	Δf_5			

cubic spline Interpolation.

Till now we have known how an interpolation polynomial of degree n can be constructed from the given set of values of function. It has been proved that when n is large, the interpolation polynomial does not provide accurate result. This illustrate in figure below.

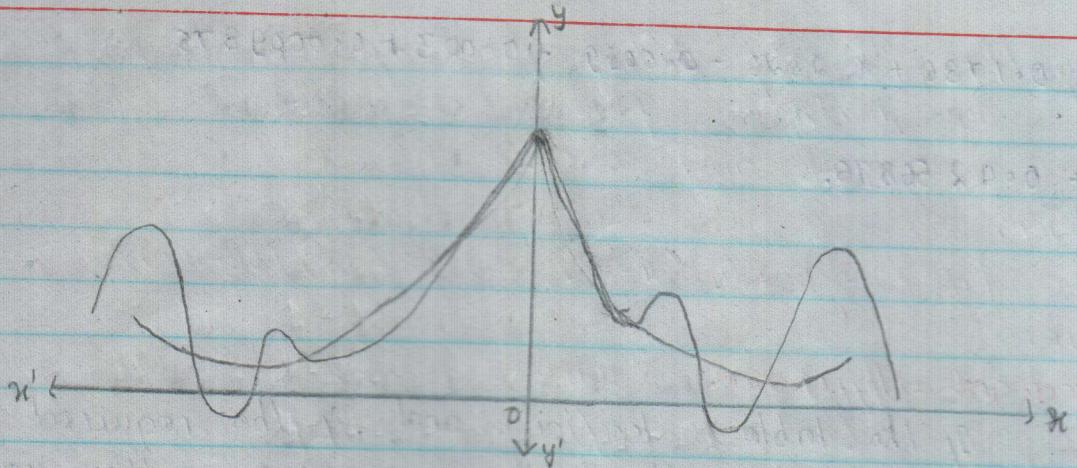


fig. Interpolation polynomial of degree 11 of the function $\frac{1}{1+x^2}$

To overcome this problem, we divide the entire range of points into sub intervals, and use local low order polynomial to interpolate each sub interval. Such polynomials are called piecewise polynomial.

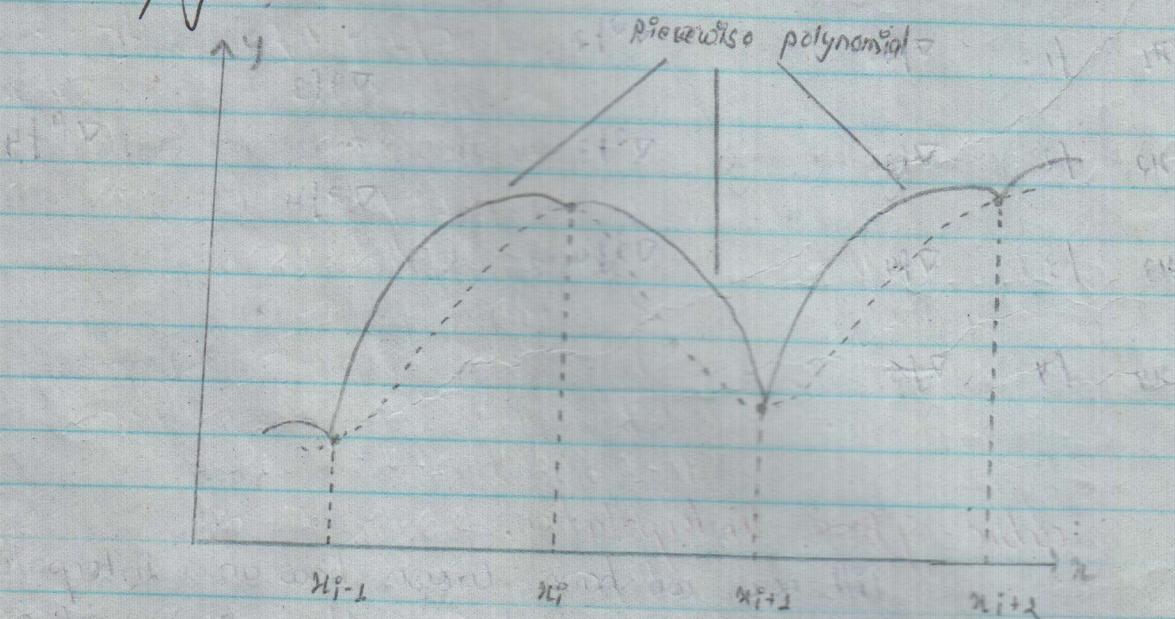


fig. piecewise polynomial interpolation.

The piecewise polynomial that prevent the discontinuities at the connecting points are called spline function (or simply splines). The connecting points are called knots or nodes.

Consider the construction of cubic spline function which would interpolate the points $(x_0, f_0), (x_1, f_1), \dots, (x_n, f_n)$. The cubic spline $S(x)$ consists of $(n-1)$ cubics corresponds to $(n-1)$ sub intervals. If we denote such cubics by $s_i(x)$ then,

$$s(x) = s_i(x)$$

where $i = 1, 2, \dots, n$.

These cubics must satisfy the following condition.

- ① $s(x)$ must interpolate at all the points x_0, x_1, \dots, x_n i.e. for each i ,

$$s(x_i) = f_i \quad \text{--- (1)}$$

- ② The function values must be equal at all the interior knots.

$$s(x_i) = s_{i+1}(x_i) \quad \text{--- (2)}$$

- ③ The first derivatives at the interior knots must be equal.

$$s'_i(x_i) = s'_{i+1}(x_i) \quad \text{--- (3)}$$

- ④ The second derivatives at the interior knots must be equal.

$$s''_i(x_i) = s''_{i+1}(x_i) \quad \text{--- (4)}$$

- ⑤ The second derivative at the end points are zero.

$$s''(x_0) = s''(x_n) = 0 \quad \text{--- (5)}$$

Step 1:

Let us consider the second derivative. Since $s(x)$ is a cubic function, its second derivative $s''(x)$ is a straight line. This straight line can be represented by a first order lagrange interpolation polynomial. Since the line passes through the points $(x_i, s''(x_i))$ and $(x_{i-1}, s''(x_{i-1}))$, we have

$$s''(x) = s''(x_{i-1}) \frac{x - x_i}{x_{i-1} - x_i} + s''(x_i) \frac{x - x_{i-1}}{x_i - x_{i-1}} \quad \text{--- (6)}$$

The unknown $s''(x_{i-1})$ and $s''(x_i)$ are to be determined.

for the sake of simplicity, let us denote,

$$S_i''(x_{i-1}) = a_{i-1}$$

$$S_i''(x_i) = a_i$$

$$x - x_i = v_i$$

$$x - x_{i-1} = v_{i-1}$$

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

Then equation (6) becomes,

$$\text{along similar } S_i''(x) = a_{i-1} \frac{v_i}{-h_i} + a_i \frac{v_{i-1}}{h_i}$$

$$= \frac{a_i v_{i-1} - a_{i-1} v_i}{h_i} \quad \dots \quad (7)$$

Step 2:

Now we can obtain $S_i(x)$ by integrating eqn (7)
hence, thus

$$S_i(x) = \frac{a_i v_{i-1}^3 - a_{i-1} v_i^3}{6h_i} + c_1 x + c_2 \quad \dots \quad (8)$$

where c_1 and c_2 are constants of integration.

The linear part $c_1 x + c_2$ can be expressed as,

$$b_1 (x - x_{i-1}) + b_2 (x - x_i)$$

with the suitable choice of b_1 and b_2 .

Therefore,

$$c_1 x + c_2 = b_1 (x - x_{i-1}) + b_2 (x - x_i)$$

Then equation (8) becomes,

$$S_i(x) = \frac{a_i v_{i-1}^3 - a_{i-1} v_i^3}{6h_i} + b_1 (v_{i-1}) + b_2 (v_i) \quad \dots \quad (9)$$

Now, we must determine the coefficient b_1 and b_2 .
 We know, by condition 9,

$$g(x_i) = f_i \text{ and}$$

$$S(2^q-1) = f^q - 1$$

At $x = x_i^o$

$$u_1^o = 0$$

$$u_{i_1}^o = h_i$$

and,

$$f_1 = \frac{q_1 h_1^2}{\delta} + b_1 h_1$$

$$\therefore \delta_s = \frac{f_i}{h_i^o} - \frac{q_i h_i^o}{6} \quad \dots \quad (10)$$

Similarly,

$$\text{at } x = kp - 1$$

$$U_I^o = -k_I^o$$

$$4j^o - 1 = \emptyset$$

and,

$$f_{g-1} = \frac{q_{i-1} b_i^g - b_g h_i}{6}$$

$$b_2 = -\frac{f_{i-1}^o}{h_i^o} + \frac{g_{i-1}^o h_i}{6} \quad \dots \quad (11)$$

Substituting for b_1 and b_2 in eqn ⑨ & solving we get.

$$S_i^o(x) = \frac{q_{i-1}}{6h_i} \cdot (h_i^3 v_i^3 - v_i^3) + \frac{q_i}{6h_i} \cdot (v_{i-1}^3 - h_i^3 v_{i-1}) \\ + \frac{1}{h_i} (f_i v_{i-1} - f_{i-1} v_i) \quad \dots \quad (12)$$

Equation 12 has only two unknowns q_{i-1} and q_i

Step-4

The final step is to evaluate these constants. This can be done by invoking the conditions

$$S_i'(x_i^*) = S_{i+1}'(x_i^*)$$

Differentiating eqⁿ (13) we get

$$\text{or } S_i'(x) = \frac{q_{i-1}}{6h_i} (f_{i-1} - 3f_i + f_{i+1}) + \frac{q_i}{6h_i} (3f_{i-1} - 6f_i + f_{i+1}) + \frac{1}{h_i} (f_i - f_{i-1})$$

Setting $x = x_i^*$

$$S_i'(x_i^*) = \frac{q_{i-1} h_i}{6} + \frac{q_i h_i}{6} - \frac{f_i - f_{i-1}}{h_i}$$

Similarly,

$$S_{i+1}'(x_i^*) = -\frac{q_i h_{i+1}}{6} - \frac{q_{i+1} h_{i+1}}{6} + \frac{f_{i+1} - f_i}{h_{i+1}}$$

$$\text{Since } S_i'(x_i^*) = S_{i+1}'(x_i^*)$$

we have,

$$\begin{aligned} & h_i q_{i-1} + 2(h_i + h_{i+1}) q_i + h_{i+1} q_{i+1} \\ \text{or } & = 6 \left[\frac{f_{i+1} - f_i}{h_{i+1}} - \frac{f_i - f_{i-1}}{h_i} \right] \quad \dots \dots \quad (13) \end{aligned}$$

Here, since the 1st derivatives at the end points are zero, we get,

$$q_0 = q_n = 0$$

Also the system of $n-1$ equations contained in eqⁿ (13) can be expressed as,

$$\left[\begin{array}{ccccc} 2(h_1 + h_2) & h_2 & 0 & \cdots & 0 & 0 \\ h_2 & 2(h_2 + h_3) & h_3 & \cdots & 0 & 0 \\ \vdots & \vdots & \ddots & \cdots & h_{n-2} & 2(h_{n-2} + h_{n-1}) h_{n-1} \\ \vdots & \vdots & & & \vdots & \\ 0 & 0 & 0 & \cdots & 0 & h_{n-1} + 2(h_{n-1} + h_n) \end{array} \right]$$

$$\begin{bmatrix} q_1 \\ q_2 \\ \vdots \\ q_{n-1} \end{bmatrix} = \begin{bmatrix} D_2 \\ D_2 \\ \vdots \\ D_{n-1} \end{bmatrix} \quad (14)$$

where,

$$q_i = 6 \left[\frac{f_{p+1} - f_p}{h_{i+1}} - \frac{f_i - f_{i-1}}{h_i} \right]$$

$$h_i = x_p - x_{p-1}$$

$$i = 1, 2, 3, \dots, n-1$$

Q1 Given the data points, estimate the function value f at $x=7$ using cubic spline

x^0	0	1	2
f^0	4	9	16
f^1	2	3	4

Sol:

$$\text{Here } h_1 = x_2 - x_0$$

$$h_2 = x_2 - x_1$$

$$= 9 - 4$$

$$= 16 - 9$$

$$= 8$$

$$f_0 = 2, \quad f_1 = 3, \quad f_2 = 4$$

Therefore,

$$h_1 q_0 + 2(h_1 + h_2)q_1 + h_2 q_2 = 6 \left[\frac{f_2 - f_1}{h_2} - \frac{f_1 - f_0}{h_1} \right]$$

we know that,

$$q_0 = q_2 = 0 \text{ Thus}$$

$$2(h_1 + h_2)q_1 = 6 \left[\frac{f_2 - f_1}{h_2} - \frac{f_1 - f_0}{h_1} \right]$$

$$\text{or } 2[5+7]q_1 = 6 \left[\frac{4-3}{7} - \frac{8-9}{8} \right]$$

$$\text{or, } 2 \times 12 q_1 = 6 \left[\frac{1}{7} - \frac{1}{8} \right]$$

$$\text{or, } \frac{24}{6} q_1 = \left[\frac{15-7}{35} \right]$$

$$\text{or, } 4q_1 = -\frac{1-2}{35}$$

$$\therefore q_1 = \frac{-2}{24 \times 35}$$

$$= -\frac{1}{70}$$

$$= -0.0143$$

Since $n=3$, there are two cubic splines namely,

$$S_1(x) : x_0 \leq x \leq x_1$$

$$S_2(x) : x_1 \leq x \leq x_2$$

the target point $x=7$ is in the domain of $S_1(x)$ and therefore we need to use only $S_1(x)$ for estimate.

$$\therefore S_1(x) = q_1 \left[(x_0^3 - h_1^2 U_0) \right] + \frac{1}{h_1} \left[f_1 U_0 - f_0 U_1 \right]$$

and..

$$U_0 = x - x_0$$

$$= 7 - 4$$

$$= 3$$

$$U_1 = x - x_1$$

$$= 7 - 9$$

$$= -2$$

f_{01}

$$\begin{aligned}
 S_1(7) &= \frac{-0.0143 (3^3 - 5^2 \cdot 3)}{6 \times 5} + \frac{1}{5} [3 \times 3 - 2 \times 2] \\
 &= -0.0143 \frac{(27 - 75)}{30} + \frac{1}{5} (9 + 4) \\
 &= 0.02288 + 2.6 \\
 &= 2.62288 //
 \end{aligned}$$

Hence $y = 2.62288$.

Q, find the cubic splines for the following table of values.

$x = 1$	2	3
$y = -6$	-1	16

Hence evaluate $y(1.5)$ and $y'(2)$.

Sol, Here we have,

$$\begin{aligned}
 h_0 &= 1, \quad h_1 = 2, \quad h_2 = 3 \\
 h_1 &= h_2 - h_0 \quad h_2 = h_2 - h_1 \\
 &= 2 - 1 \quad = 8 - 2 \\
 &= 1 \quad = 1.
 \end{aligned}$$

$$f_0 = -6 \quad f_1 = -1 \quad f_2 = 16$$

Therefore,

$$h_1 q_0 + 2(h_1 + h_2) q_1 + h_2 q_2 = 6 \left[\frac{f_2 - f_1}{h_2} - \frac{f_1 - f_0}{h_1} \right]$$

$$\text{or, } 1 \times 0 + 2(1 + 3) q_1 + 1 \times 0 = 6 \left[\frac{16 - -1}{1} - \frac{(-1) - (-6)}{1} \right]$$

$$\text{or, } 4q_1 = 6 [13 - 5]$$

$$\text{or } 4q_1 = 6 [22]$$

$$\text{or } q_1 = 18 //$$

78

Since $n=3$, there are two cubic splines only.

$$S_1(x) = u_0 \leq x \leq x_1$$

$$S_2(x) = x_1 \leq x_0 \leq x_2$$

Here the given point $x=10.5$ is in the domain of $S_1(x)$ & therefore we use only $S_1(x)$ for estimate,

$$\begin{aligned} h_0 &= x - x_0 & v_1 &= x - x_1 \\ &= 10.5 - 1 & &= 10.5 - 9 \\ &= 0.05 & &= -0.05 \end{aligned}$$

then

$$\begin{aligned} y(10.5) &= \frac{1}{6} [0.05^3 - 1^2 \cdot 0.005] + \frac{1}{1} [-1 \cdot 0.05 - (-6) \times (-0.05)] \\ &= -10.125 + (-3.05) \\ &= -40.625 \end{aligned}$$

$$S_1(x) = \frac{a_1 (u_0^3 - h_1^2 u_0)}{6 h_1} + \frac{s_1 \times v_1}{h_1} [f(x_{u_0}) - f(x_{u_1})]$$

$$S_1(2) = 11$$

Least Square Approximation.

The process of establishing the relationship between dependent variable & independent variable in the form of mathematical equation is known as curve fitting (regression). Suppose the value of y for the different values of x are given. If we want to know effect of x on y then we may write a functional relationship

$$y = f(x)$$

Here y is called dependent variable.
 x is independent variable.

The relationship may be linear or non linear as shown in the figure below-

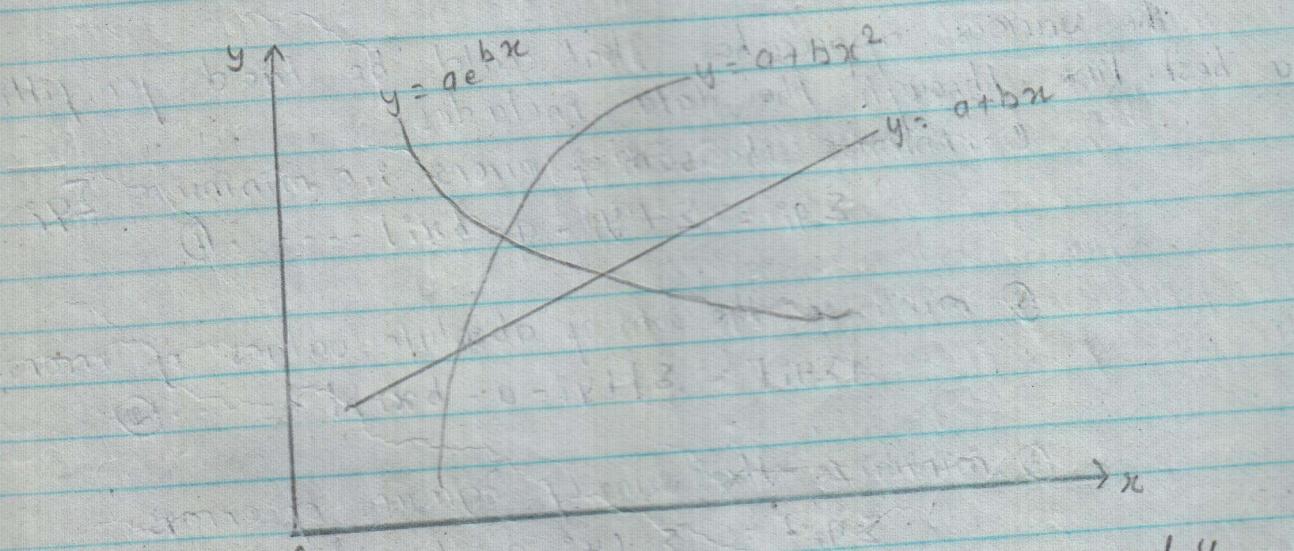


Fig:- various relationship between x and y .

Fitting Linear equation.

Let $y = a + bx = f(x)$ be the mathematical equation for a straight line.

From this equation we know that a is the intercept of the line and b its slope.

Consider a point (x_i, y_i) as shown in the figure below.

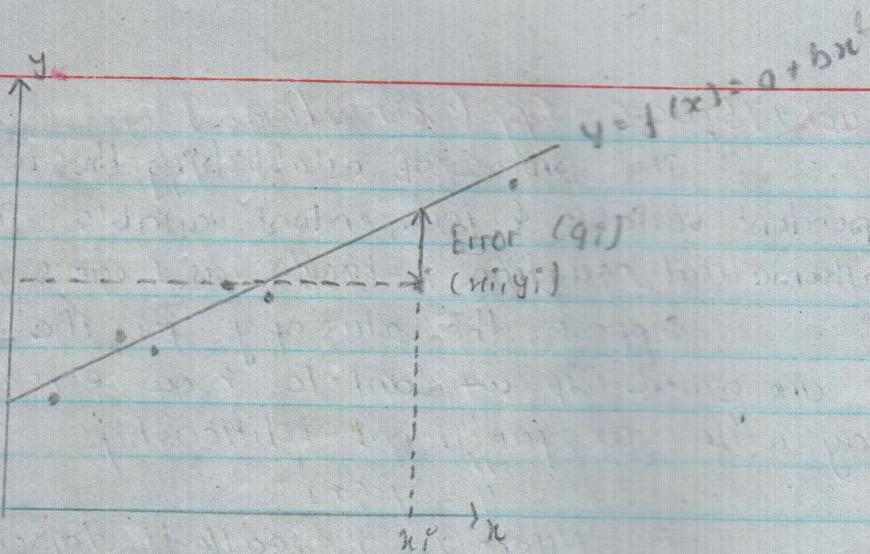


fig: Scatter diagram.

Here if q_i is the error, then

$$q_i = y_i - f(x_i)$$

The various approaches that could be tried for fitting a best line through the data includes

① minimize the sum of errors. i.e minimize Σq_i

$$\Sigma q_i = \Sigma (y_i - a - bx_i) \dots \textcircled{1}$$

② minimize the sum of absolute values of errors.

$$\sum |q_i| = \sum |(y_i - a - bx_i)| \dots \textcircled{2}$$

③ minimize the sum of squares of errors.

$$\Sigma q_i^2 = \Sigma (y_i - a - bx_i)^2 \dots \textcircled{3}$$

from above equations, only eqn ③ guarantees a unique line. Hence the technique of minimizing the sum of squares of errors is known as least square regression.

Least square regression.

Let the sum of squares of individual errors be expressed as

$$\begin{aligned} Q &= \sum_{i=1}^n q_i^2 \\ &= \sum_{i=1}^n [y_i - f(x_i)]^2 \\ &= \sum_{i=1}^n [y_i - a - bx_i]^2 \quad \dots \dots \dots \textcircled{1} \end{aligned}$$

In the method of least squares we choose a and b such that Q is minimum. Since Q depends on a and b , a necessary condition for Q to be minimum is,

$$\frac{\partial Q}{\partial a} = 0 \quad \text{and} \quad \frac{\partial Q}{\partial b} = 0$$

Then,

$$\frac{\partial Q}{\partial a} = -2 \sum_{i=1}^n [y_i - a - bx_i] = 0$$

$$\frac{\partial Q}{\partial b} = -2 \sum_{i=1}^n x_i [y_i - a - bx_i] = 0$$

Thus,

$$\sum y_i = na + b \sum x_i$$

and

$$\sum x_i y_i = a \sum x_i + b \sum x_i^2$$

These are called normal equations, solving for a and b we get

$$b = \frac{n \sum x_i y_i - \sum x_i \sum y_i}{n \sum x_i^2 - (\sum x_i)^2}$$

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

where \bar{x} and \bar{y} are the average of x values and y values respectively.

Q11 Fit a straight line to the following set of data.

x_i	1	2	3	4	5
y_i	3	4	5	6	8

Soln

x_i	y_i	x_i^2	$x_i y_i$
1	3	1	3
2	4	4	8
3	5	9	15
4	6	16	24
5	8	25	40
sum =	15	55	90

$$\text{Then, } b = \frac{n \sum x_i y_i - \sum x_i \sum y_i}{n \sum x_i^2 - (\sum x_i)^2}$$

$$= \frac{5 \times 90 - 15 \times 26}{5 \times 55 - (15)^2}$$

$$= \frac{450 - 390}{275 - 225}$$

$$= \frac{60}{50}$$

$$= 1.2$$

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

$$= \frac{26}{5} - 1.2 \times \frac{15}{5}$$

$$= 5.2 - 3.6$$

$$= 1.6$$

Thus the linear equation is,

$$y = 1.6 + 1.02x$$

The regression line along with the above data is shown in the figure below.

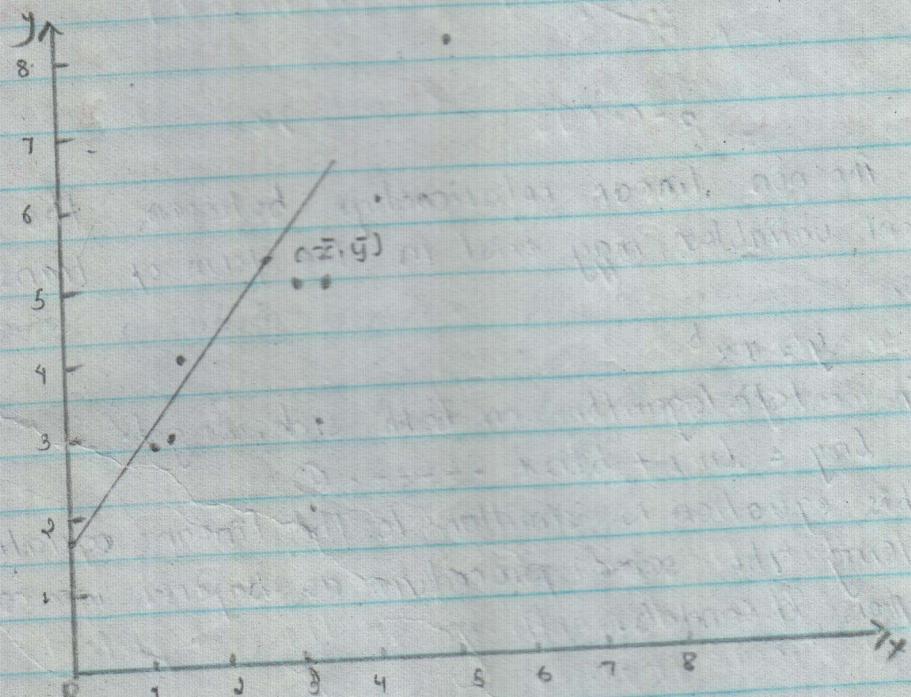


Fig.: plot of data and regression line of above example.

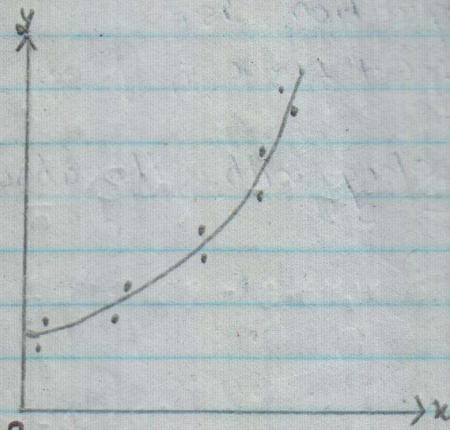
Algorithm for linear regression

1. Read data values.
2. Compute the sum of powers & products i.e. $\sum x_i$, $\sum y_i$, $\sum x_i^2$, $\sum x_i y_i$.
3. Check whether the denominator for the equation of b is 0.
4. Compute b and a.
5. Printout the equation.
6. Interpolate data if required.

84

Fitting Transcendental Equation (exponential) :-

The relationship between dependent & independent variables is not always linear, as shown in the figure below



The non linear relationship between the dependent & independent variables may exist in the form of transcendental equation as,

$$y = ax^b$$

If we take logarithm on both sides, we get.

$$\ln y = \ln a + b \ln x \quad \dots \dots \dots \textcircled{1}$$

This equation is similar to the linear equation and therefore using the same procedure as before, we can evaluate the parameters a and b .

$$b = \frac{n \sum \ln x_i \ln y_i - \sum \ln x_i \sum \ln y_i}{n \sum (\ln x_i)^2 - (\sum \ln x_i)^2}$$

$$\ln a = R = \frac{1}{n} [\sum \ln y_i - b \sum \ln x_i]$$

$$a = e^R$$

There is another form of non linear model known as saturation growth rate, equation as shown below;

$$P = \frac{k_1 t}{k_2 + t}$$

This can be linearized by taking inversion of the terms,

$$\text{i.e., } \frac{t}{P} = \frac{K_2}{K_1} \cdot \frac{1}{t} + \frac{1}{K_1}$$

This is again similar to linear equation

$y = a + bx$ where,

$$y = \frac{1}{P}, \quad x = \frac{1}{t}, \quad a = \frac{1}{K_1}, \quad b = \frac{K_2}{K_1}$$

Q1 Given the data table,

x	1	2	3	4	5
y	0.5	2	4.5	8	12.5

fit a power function model of the form $y = a x^b$

Soln,

various quantities required for the above equations are tabulated below,

x_i	y_i	$\ln x_i$	$\ln y_i$	$(\ln x_i)^2$	$(\ln x_i)(\ln y_i)$
1	0.5	0	-0.693	0	0
2	2	0.693	0.693	0.48	0.48
3	4.5	1.398	1.504	1.897	1.652
4	8	2.207	2.079	4.891	2.881
5	12.5	2.690	2.526	6.989	4.064
Sum =	15	4.786	6.109	6.197	9.077

Now,

$$b = \frac{n \sum \ln x_i \ln y_i - \sum \ln x_i \sum \ln y_i}{n \sum (\ln x_i)^2 - (\sum \ln x_i)^2}$$

$$= \frac{5 \times 9.077 - 4.786 \times 6.109}{5 \times (6.197) - (4.786)^2}$$

$$= \frac{45.385 - 29.243}{14.5 - 22.98}$$

$$= \frac{16.141}{2.8498} = \frac{16.141}{8.069}$$

$$= 0.616 = b$$

Then,

$$q = e^b$$

$$= e^{\frac{1}{n} (\sum \ln y_i - b \sum \ln x_i)}$$

$$= e^{\frac{1}{5} (6.109 - 0.616 \times 4.787)}$$

$$= e^{-0.693}$$

$$= 0.511$$

Now, we know

$$y = q n^b$$

$$= 0.5 x^2$$

Note that the data have been derived from the eqn
 $y = n^2/2$. The discrepancy in the computed coefficients
is due to roundoff error.

Q1. The temperature of a metal strip was measured at various intervals during heating and the values are given in a table below

Time t (min) 1 2 3 4

Temp T ($^{\circ}\text{C}$) 70 83 100 124

If the relationship between the temperature T and time t is of the form

$T = b t^{1/4} + q$ estimate the temperature at $t = 6$ min.

Soln.

The relationship between temperature & time can be written as,

$$y = b f(x) + a$$

This is similar to the linear eqn except that the variable x is replaced by function $f(x)$ therefore we can solve for the parameters a & b by replacing x_i by $f(x_i)$, $\sum x_i$ by $\sum f(x_i)$.

Fitting a polynomial function.

consider a polynomial of degree $m-1$, $y = 0$

$$y = a_1 + a_2 x + a_3 x^2 + \dots + a_m x^{m-1} = f(x)$$

If there is a set of data for x and y values, then the sum of squares of errors is given by,

$$\theta = \sum_{i=1}^n [y_i - f(x_i)]^2$$

Since $f(x)$ is polynomial and contains coefficients a_1, a_2, a_3, \dots etc., we have to estimate all the m coefficients.

following m equations solve these coefficients.

$$\frac{\partial \theta}{\partial a_1} = 0$$

$$\frac{\partial \theta}{\partial a_2} = 0$$

$$\frac{\partial \theta}{\partial a_m} = 0$$

Consider a general term,

$$\frac{\partial \theta}{\partial a_j} = -2 \sum_{i=1}^n [y_i - f(x_i)] \frac{\partial f(x_i)}{\partial a_j} = 0$$

$$\frac{\partial f(x_i)}{\partial a_j} = x_i^{j-1}$$

thus we have,

$$\sum_{i=1}^n [y_i - f(x_i)] x_i^{j-1} = 0, j = 1, 2, \dots, m.$$

$$\sum_{i=1}^n [y_i x_i^{j-1} - x_i^{j-1} f(x_i)] = 0$$

Substituting for $f(x_i)$

$$\sum_{j=1}^n x_i^{j-1} [q_1 + q_2 x_i + q_3 x_i^2 + \dots + q_m x_i^{m-1}] = \sum_{j=1}^n y_i x_i^{j-1}$$

These m equations ($j = 1, 2, 3, \dots, m$) and each summation goes for $i = 1$ to n

$$q_1 n + q_2 \sum x_i + q_3 \sum x_i^2 + \dots + q_m \sum x_i^{m-1} = \sum y_i$$

$$q_1 \sum x_i + q_2 \sum x_i^2 + q_3 \sum x_i^3 + \dots + q_m \sum x_i^m = \sum y_i x_i$$

$$q_1 \sum x_i^{m-1} + q_2 \sum x_i^m + q_3 \sum x_i^{m+1} + \dots + q_m \sum x_i^{2m-2} = \sum y_i x_i^{m-1}$$

The set of m equations can be represented in matrix notation as follows,

$$CA = B$$

where,

$$C = \begin{bmatrix} n & \sum x_i & \sum x_i^2 & \sum x_i^{m-1} \\ \sum x_i & \sum x_i^2 & \sum x_i^3 & \sum x_i^m \\ \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \\ \sum x_i^{m-1} & \sum x_i^m & \dots & \sum x_i^{2m-2} \end{bmatrix}$$

$$A = \begin{bmatrix} q_1 \\ q_2 \\ \vdots \\ q_m \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} \sum y_i \\ \sum y_i x_i \\ \vdots \\ \sum y_i x_i^{m-1} \end{bmatrix}$$

Temp
y_i Fit a second order polynomial to the data in the table below.

x	1.0	2.0	3.0	4.0
---	-----	-----	-----	-----

y	6.0	11.6	18.0	-27.0
---	-----	------	------	-------

#

SOLN

The order of polynomial is too and therefore we will have three simultaneous eqn as shown below,

$$q_1 n + q_2 \sum x_i^1 + q_3 \sum x_i^{1+2} = \sum y_i^1$$

$$q_1 \sum x_i^1 + q_2 \sum x_i^{1+2} + q_3 \sum x_i^{1+3} = \sum x_i^1 y_i^1$$

$$q_1 \sum x_i^{1+2} + q_2 \sum x_i^{1+3} + q_3 \sum x_i^{1+4} = \sum x_i^2 y_i^1$$

The term of powers & product can be evaluated in tabular form as shown below.

x^1	y^1	x^2	x^3	x^4	xy^1	x^2y^1
1	6	1	1	1	6	6
2	11	4	8	16	22	44
3	18	9	27	81	84	162
4	27	16	64	256	108	432
sum =	10	62	100	354	190	644

Substituting these values we get,

$$4q_1 + 10q_2 + 30q_3 = 62$$

$$10q_1 + 30q_2 + 100q_3 = 190$$

$$30q_1 + 100q_2 + 354q_3 = 644$$

Solving we get

$$q_1 = 3, q_2 = 2, q_3 = 1$$

Therefore the best square quadratic polynomial is,

$$y = 3 + 2x + x^2$$

Solution of Linear Algebraic Equation,

A linear equation involving two variables x and y has the standard form,

$$ax + by = c$$

where a, b and c are real numbers and $a \neq b$ cannot be both equal to zero. ($a, b \neq 0$)

Analysis of linear equation is important because mathematical model of many of the real world problems are linear.

In practice linear equations occur in more than two variables. A linear eqn with n variables has the form,

$$a_1x_1 + a_2x_2 + a_3x_3 + \dots + a_nx_n = b.$$

where a_i ($i = 1, 2, 3, \dots, n$) are real numbers and atleast one of them is not zero. The main concern here is to solve for x_i ($i = 1, 2, 3, \dots, n$), given the values of a_i and b .

If we need a unique soln of an eqn with n variables, then we need a set of n such independent equations which is known as system of simultaneous equations [or simply system of equations]. A system of n linear equation is represented as

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

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

$$\vdots$$

$$\vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n$$

In matrix form the above equation can be expressed as

$$Ax = b$$

where A is an $n \times n$ matrix, b is an n vector & x is a vector of n unknowns.

Existence of solution.

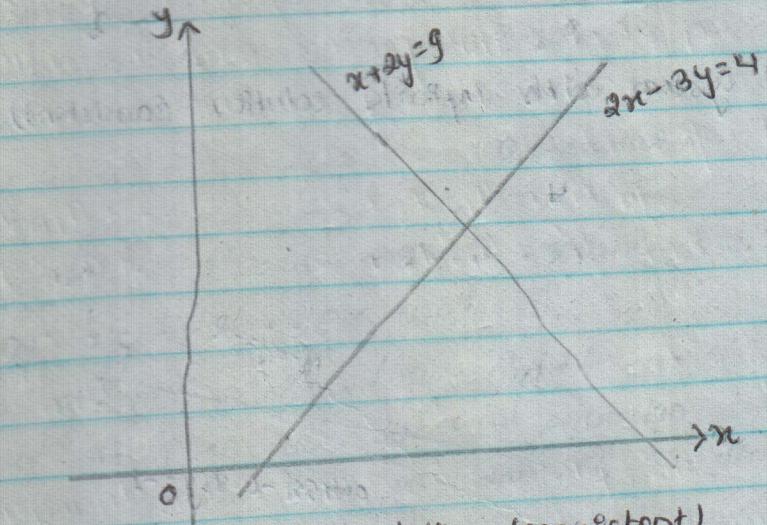
Given an arbitrary system of equation. It is difficult to say whether the system has a soln or not. There are 4 possibilities.

① System has a unique solution.

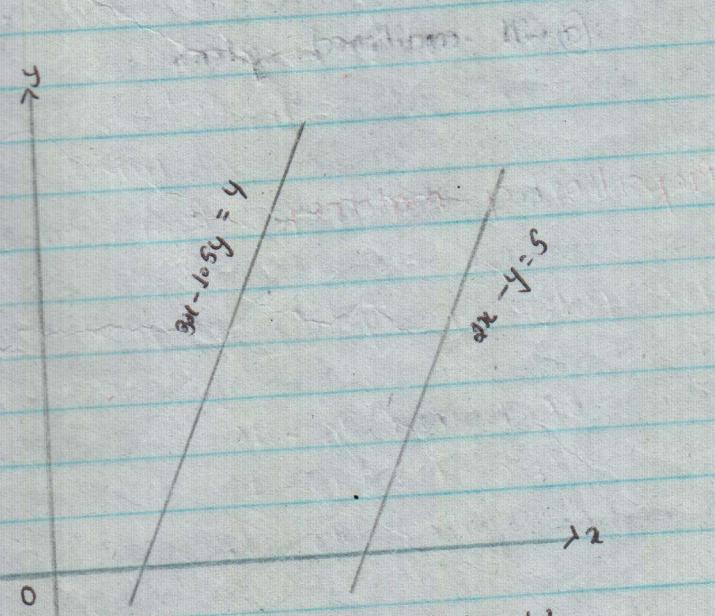
② System has no solution.

③ System has a solution - but not a unique one. [Infinite solns]

④ System is illconditioned.

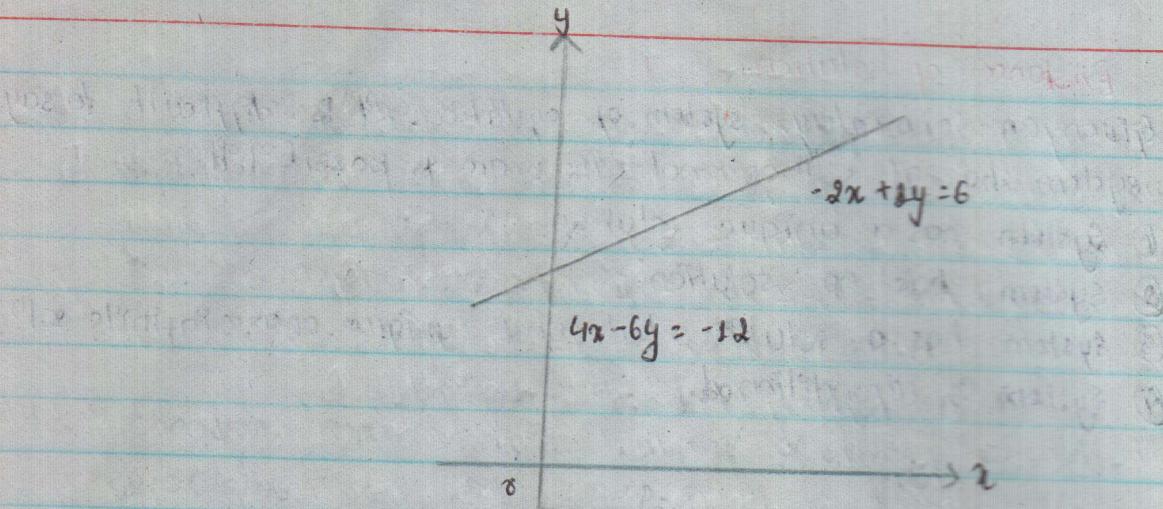


⑤ System with unique solution (consistent)

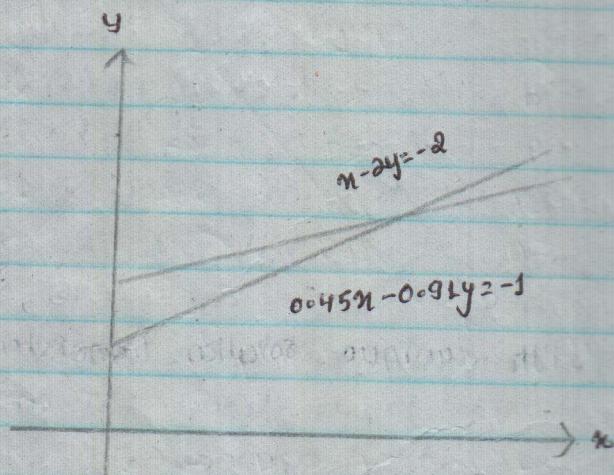


⑥ System with no solution (inconsistent)

94



① System with infinite solution (consistent)



② All -conditioned system.

Properties of matrices.

Gaussian Elimination method:

Gauss elimination method reduces the system of equations to an equivalent upper triangular system which can be solved by backward substitution.

Let us consider a general set of n equation and n unknown variable.

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

$$a_{21}x_1 + a_{22}x_2 + \dots - a_{2n}x_n = b_2$$

$$\dots \dots \dots \dots \dots \dots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots - a_{nn}x_n = b_n$$

Let us assume that a solution exist and that is unique. Following algorithms are used.

① Arrange equations such that $a_{11} \neq 0$.

② Eliminate x_1 from all but ^{except} 1st equation as follows.

③ Normalize the first equation by dividing it by a_{11} .

④ Subtract from the second equation a_{21} times the normalized 1st equation. The result is

$$\left[a_{21} - a_{21} \frac{a_{11}}{a_{11}} \right] x_1 + \left[a_{22} - a_{21} \frac{a_{12}}{a_{11}} \right] x_2 + \dots = b_2 - a_{21} \frac{b_1}{a_{11}}$$

$$\text{we can see that } a_{21} - a_{21} \frac{a_{11}}{a_{11}} = 0$$

Thus the resultant equation does not contain x_1 , the new second equation is

$$0 + a_{22}'x_2 + a_{23}'x_3 + \dots - a_{2n}'x_n = b_2'$$

⑤ Similarly subtract from the third equation a_{31} times the normalized 1st equation. The result will be

$$0 + a_{32}'x_2 + \dots - a_{3n}'x_n = b_3'$$

continuing this process we get,

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

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

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

③ eliminate x_2 from the third to the last equation in the new set, again we assume that $a_{22} \neq 0$

④ subtract from the third equation a_{22}' times the normalized second equation.

⑤ subtract from the fourth equation a_{32}' times the normalized second equation & do on.

This process will continue till the last equation contains only one unknown, namely x_n . The final form is

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

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

$$a_{nn}^{(n-1)}x_n = b_n^{n-1}$$

This process is called triangulation. The number of primes indicate the number of times the coefficient have been modified.

⑥ obtain the solution by back substitution. The solution is as follows,

$$x_n = \frac{b_n^{n-1}}{a_{nn}^{n-1}}$$

This can be substituted back in the $n-1^{\text{th}}$ eq to obtain the solution for $x_{n-1}, x_{n-2}, \dots, x_1$.

Q1) Solve the following system of equations using Gauss elimination method

$$3x_1 + 6x_2 + x_3 = 16$$

$$2x_1 + 4x_2 + 3x_3 = 13$$

$$x_1 + 3x_2 + 2x_3 = 9.$$

Solution,

The given equations are,

$$3x_1 + 6x_2 + x_3 = 16$$

$$2x_1 + 4x_2 + 3x_3 = 13$$

$$x_1 + 2x_2 + 2x_3 = 9$$

Now,

the first normalized equation is

$$\frac{2}{3}x_1 + \frac{4}{3}x_2 + \frac{3}{3}x_3 = \frac{13}{3}$$

$$x_1 + 2x_2 + \frac{1}{3}x_3 = \frac{13}{3}$$

Multiplying this equation by a_{31} i.e. 2 we get,

$$2x_1 + 4x_2 + \frac{2}{3}x_3 = \frac{26}{3}$$

Subtracting this equation from equation (2) we get

$$(2x_1 + 4x_2 + 3x_3) - (2x_1 + 4x_2 + \frac{2}{3}x_3) = 13 - \frac{26}{3}$$

$$0 + 0 + 3x_3 - \frac{2}{3}x_3 = \frac{39 - 26}{3}$$

$$\text{or, } \frac{7x_3}{3} = \frac{7}{3}$$

$$\therefore x_3 = 1,$$

for 3rd eqn, multiplying 1st normalized equation by a_{31} i.e. 1 we get

$$x_1 + 2x_2 + \frac{1}{3}x_3 = \frac{16}{3}$$

Subtracting this equation from eqn (3) we get

$$(x_1 + 2x_2 + 3x_3) - (x_1 + 2x_2 + \frac{1}{3}x_3) = 9 - \frac{16}{3}$$

$$\text{or, } x_2 + \frac{5x_3}{3} = \frac{11}{3}$$

Now we know the value of x_3 , so by back substitution, we get

$$x_2 + \frac{5 \cdot 1}{3} = \frac{11}{3}$$

$$x_2 = \frac{11}{3} - \frac{5}{3}$$

98

$$\begin{aligned}H_2 &= \frac{11 - 15}{3} \\&= -\frac{4}{3} \\&= 2\frac{2}{3}\end{aligned}$$

and,

$$3x_1 + 6x_2 + x_3 = 16$$

$$\text{or } 3x_1 + 6x_2 + 1 = 16$$

$$\text{or } 3x_1 + 13 = 16$$

$$\text{or } 3x_1 = 16 - 13$$

$$\text{or } 3x_1 = 3$$

$$\therefore x_1 = 1$$

Therefore

$$x_1 = 1$$

$$x_2 = 4$$

$$x_3 = 1$$

Gauss Elimination with pivoting.

In basic gaussian elimination method, the elimination step where $i=j$ is known as pivot element. Each row is normalized by dividing the coefficient of that row by its pivot element. If a_{kk} is zero, k^{th} row cannot be normalized and the procedure fails. One way to overcome this problem is to interchange this row with another row below it which does not have a zero element and have largest (absolute value) coefficient in that position.

The procedure of re-ordering involves the following steps.

- ① Search & locate the largest absolute value among the coefficients in the first column.
- ② Exchange the first row with the row containing that element.
- ③ Then eliminate the 1st variable in the second equation as explained earlier.
- ④ When the second row becomes pivot row, search for the coefficients in the 2nd column from the 2nd row to the n^{th} row and locate the largest coefficient. Exchange the 2nd row with the row containing largest coefficient.
- ⑤ Continue this process until $n-1$ unknown are eliminated.

Q/H

Solve the following system of equations using partial pivoting technique,

$$2x_1 + 2x_2 + x_3 = 6$$

$$4x_1 + 2x_2 + 3x_3 = 4$$

$$x_1 - x_2 + 2x_3 = 0$$

SOL,

The given equations are,

$$2x_1 + 2x_2 + x_3 = 6$$

$$4x_1 + 2x_2 + 3x_3 = 4$$

100

$$x_1 - x_2 + x_3 = 0$$

Here in these three equations, the highest coefficient of x_1 is in second equation, so interchanging second equation with first equation we get,

$$4x_1 + 2x_2 + 3x_3 = 4$$

$$2x_1 + 2x_2 + x_3 = 6$$

$$x_1 - x_2 + x_3 = 0$$

Now normalized 1st equation we get,

$$\frac{4}{4}x_1 + \frac{2}{4}x_2 + \frac{3}{4}x_3 = \frac{4}{4}$$

$$\text{or } x_1 + \frac{1}{2}x_2 + \frac{3}{4}x_3 = 1,$$

Multiplying this equation by 2 we get,

$$2x_1 + x_2 + \frac{3}{2}x_3 = 2.$$

Subtracting this equation from second equation we get,

$$(2x_1 + 2x_2 + x_3) - (2x_1 + x_2 + \frac{3}{2}x_3) = 6 - 2$$

$$\text{or, } 0 + x_2 + x_3 - \frac{3}{2}x_3 = 4$$

$$\text{or } x_2 - \frac{x_3}{2} = 4$$

for third equation multiplying the normalized 1st eqn by 1 we get,

$$x_1 + \frac{1}{2}x_2 + \frac{3}{4}x_3 = 1,$$

Subtracting this eqn from third equation we get,

$$(x_1 - x_2 + x_3) - (x_1 + \frac{1}{2}x_2 + \frac{3}{4}x_3) = 0 - 1$$

$$\text{or, } -x_2 - \frac{1}{2}x_2 + x_3 - \frac{3}{4}x_3 = -1.$$

$$\text{or, } -\frac{8H_2}{2} + \frac{H_3}{4} = -1$$

$$\text{or, } -6H_2 + H_3 = -4$$

$$\text{or, } 6H_2 - H_3 = 4.$$

The system of linear equations becomes,

$$4H_1 + 2H_2 + 3H_3 = 4$$

$$2H_2 - H_3 = 8$$

$$6H_2 - H_3 = 4.$$

Now comparing the coefficients of H_2 in second & 3rd row we get the greater coefficient in 3rd row, so we interchange 2nd & 3rd row,

$$4H_1 + 2H_2 + 3H_3 = 4$$

$$6H_2 - H_3 = 4$$

$$2H_2 - H_3 = 8$$

Now Normalized second eq is

$$\frac{6}{6}H_2 - \frac{1}{6}H_3 = \frac{4}{6}$$

$$\text{or, } 6H_2 - \frac{1}{6}H_3 = \frac{4}{6}$$

Multiplying this equation by 2 we get

$$2H_2 - \frac{1}{3}H_3 = \frac{4}{3}$$

Subtracting this eq from row 3 we get

$$(2H_2 - H_3) - (2H_2 - \frac{1}{3}H_3) = 8 - \frac{4}{3}$$

$$-H_3 + \frac{1}{3}H_3 = \frac{24 - 4}{3}$$

$$-\frac{2H_3}{3} = \frac{20}{3}$$

$$2H_3 = -20$$

$$H_3 = -10$$

by back substitution, of row 2 we get

$$8H_2 - H_3 = 4$$

$$8H_2 - (-10) = 4$$

$$8H_2 + 10 = 4$$

$$8H_2 = -6$$

$$H_2 = -\frac{1}{11}$$

and,

$$4H_1 + 2H_2 + 3H_3 = 4$$

$$\text{or } 4H_1 - 2 - 3 \times 10 = 4$$

$$\text{or } 4H_1 - 2 - 30 = 4$$

$$\text{or } 4H_1 - 32 = 4$$

$$\text{or } 4H_1 = 36$$

$$\text{or } H_1 = \frac{36}{4}$$

$$= \frac{9}{11}$$

$$\therefore H_1 = \frac{9}{11}, H_2 = -\frac{1}{11}, H_3 = -10$$

$$H_2 = -\frac{1}{11}$$

$$H_3 = -10$$

Gauss-Jordan Method:

JCB

It is another method used for solving a system of linear equations. In this method a variable is eliminated from the rows below & above the pivot equation. This process thus eliminates all the off-diagonal terms producing a diagonal matrix rather than a triangular matrix. Further all rows are normalized by dividing them by their pivot elements. This is illustrated in figure as follows.

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a_{22}' & a_{23}' \\ 0 & 0 & a_{33}'' \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2' \\ b_3'' \end{bmatrix}$$

Result of Gauss elimination

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1^* \\ x_2^* \\ x_3^* \end{bmatrix} = \begin{bmatrix} b_1^* \\ b_2^* \\ b_3^* \end{bmatrix}$$

Result of Gauss-Jordan elimination

The algorithm for Gauss-Jordan method is as follows-

- ① Normalize first eqn by dividing it by its pivot element.
- ② Eliminate x_1 term from all other equations.
- ③ Now, normalize the 2nd equation by dividing it by pivot element.
- ④ Eliminate x_2 from all the equations above & below the normalized pivotal equation.
- ⑤ Repeat this process until x_n is eliminated from all but from the last equation.

⑥ The resultant b vector is the solution vector.

The Gauss-Jordan method requires approximately 20% more arithmetic operation compare to Gauss method therefore this method is rarely used.

Q Solve the system of linear equations by Gauss-Jordan method.

$$2x_1 + 4x_2 - 6x_3 = -8$$

$$x_1 + 2x_2 + x_3 = 10$$

$$2x_1 - 4x_2 - 2x_3 = -12$$

Solution,

The given equations are,

$$2x_1 + 4x_2 - 6x_3 = -8 \quad \text{--- (1)}$$

$$x_1 + 2x_2 + x_3 = 10 \quad \text{--- (2)}$$

$$2x_1 - 4x_2 - 2x_3 = -12 \quad \text{--- (3)}$$

Normalized 1st equation is,

$$x_1 + 2x_2 - 3x_3 = -4$$

Multiplying this eqⁿ by 1 we get

$$x_1 + 2x_2 - 3x_3 = -4$$

Subtracting this eqⁿ from eqⁿ (2) we get

$$(x_1 + 2x_2 + x_3) - (x_1 + 2x_2 - 3x_3) = 10 - (-4)$$

$$\text{or} \quad 0 + 0 + x_3 + 3x_3 = 14$$

$$\text{or} \quad 4x_3 = 14$$

$$x_3 = \frac{14}{4}$$

$$= 3.5,$$

Again multiplying normalized 1st eqⁿ by 2 we get,

$$2x_1 + 4x_2 - 6x_3 = -8$$

Subtracting from 3rd we get

$$\begin{aligned} -4x_2 - 4x_2 - 2x_3 + 6x_3 &= 12 - (-8) \\ \text{or, } -8x_2 + 4x_3 &= -4 \end{aligned}$$

Now, the system of linear eqⁿ becomes,

$$x_1 + 2x_2 - 3x_3 = -4 \quad \dots \textcircled{1}$$

$$x_3 = 3 \text{ p} 5 \quad \dots \textcircled{11}$$

$$-8x_2 + 4x_3 = -4 \quad \dots \textcircled{111}$$

Interchanging eqⁿ $\textcircled{11}$ & $\textcircled{3}$

$$x_1 + 2x_2 - 3x_3 = -4 \quad \dots \textcircled{1}$$

$$-8x_2 + 4x_3 = -4 \quad \dots \textcircled{11}$$

$$x_3 = 3 \text{ p} 5 \quad \dots \textcircled{111}$$

Now, dividing eqⁿ $\textcircled{11}$ by -8 to get normalized form

$$x_2 - 0.5x_3 = -0.4375$$

multiplying this eqⁿ by 2 and subtracting from eqⁿ $\textcircled{1}$

$$(x_1 + 2x_2 - 3x_3) - (2x_2 - x_3) = -4 - (-0.875)$$

$$\text{or } x_1 - 0 - 2x_3 + x_3 = -4 + 0.875$$

$$\text{or } x_1 - 2x_3 = -3.125,$$

Hence the system of equations becomes,

$$x_1 - 2x_3 = -3.125 \quad \dots \textcircled{1}$$

$$x_2 - 0.5x_3 = -0.4375 \quad \dots \textcircled{11}$$

$$x_3 = -4 \quad \dots \textcircled{111}$$

Now multiplying 3rd eqⁿ by -0.5 and adding with eqⁿ $\textcircled{2}$
we get.

$$x_2 - 0.5x_3 - (-0.5x_3) = -0.4375 - (-2)$$

$$\text{or } x_2 = 1.5625$$

10⁶

Now multiplying 3rd eqⁿ by -2 & subtracting from eq¹ ①

$$(x_1 - 2x_3) - (-2x_3) = -30125 - (-8)$$

$$\text{or, } x_1 = -110128.$$

Hence the system of linear eqⁿ becomes

$$x_1 + 0x_2 + 0x_3 = -110128$$

$$0x_1 + x_2 + 0x_3 = 105625$$

$$0x_1 + 0x_2 + x_3 = -905$$

which can be represented as

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -110128 \\ 105625 \\ -905 \end{bmatrix}$$

Hence the solution vector is

$$\begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix}$$

$$\begin{bmatrix} -110128 \\ 105625 \\ -905 \end{bmatrix}$$