Numerical Analysis : [MA214] Lecture 3

Instructor: Prof. Tony J. Puthenpurakal

Department of Mathematics Indian Institute of Technology Bombay

tputhen@math.iitb.ac.in

Recall

Let $f:[a,b] \longrightarrow \mathbb{R}$ be a function and let x_0, x_1, \cdots, x_n be distinct points in [a,b] Then there exists a unique polynomial $P_n(x)$ such that

$$P_n(x_i) = f(x_i), \forall i = 0, 1, \cdots, n$$

 $P_n(x) o$ Interpolating polynomial of f ($w.r.t. x_0, x_1, \cdots, x_n$)

we had two forms of $P_n(x)$

Lagrange's form

$$l_k(x) = \prod_{i=0, i \neq k}^{n} \frac{x - x_i}{x_k - x_i}, k = 0, 1, \dots, n$$

$$P_n(x) = \sum_{i=0}^{n} f(x_i) l_i(x)$$

Advantage of Lagrange's form:- It is simple.

Disvantage of Lagrange's Form:-

There is no way of using Lagrange form of $P_n(x)$ to determine $P_{n+1}(x)$.

Newtons form of interpolating polynomial

$$P_{n}(x) = f(x_{0}) + f[x_{0}, x_{1}](x - x_{0}) + f[x_{0}, x_{1}, x_{2}](x - x_{0})(x - x_{1}) + \cdots$$

$$\vdots + f[x_{0}, x_{1}, x_{2}, \cdots, x_{n}] \prod_{i=0}^{n-1} (x - x_{i}) + f[x_{0}, x_{1}, x_{2}, \cdots, x_{n}] \prod_{i=0}^{n-1} (x - x_{i})$$

$$P_{n}(x) = P_{n-1}(x) + f[x_{0}, x_{1}, \cdots, x_{n}] \prod_{i=0}^{n-1} (x - x_{i}) + f[x_{0},$$

Recall:- Error of interpolating polynomial

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

$$e_n(\overline{x}) = f[x_0, x_1, \dots, x_n, \overline{x}] \prod_{i=0}^n (x - x_i)$$

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

$$\Psi_{n+1}(x) = \prod_{j=0}^n (x - x_j)$$
Not in syllabus on how to minimize $\Psi_{n+1}(x)$

Osculatory interpolation

Sometimes we have

$$x_0, x_1, \dots, x_k$$

 $f(x_0), f(x_1), f(x_2), \dots, f(x_k)$
 $f'(x_0), f'(x_1), f'(x_2), \dots, f'(x_k)$

We need a polynomial p(x) such that

$$p(x_i) = f(x_i)$$
 and $p'(x_i) = f'(x_i)$, for $i = 0, 1, \dots, n$

Note:- deg $p(x) \le 2n + 1$

Example:-

$$\frac{dy}{dx} = g(x,y), y(x_0) = y_0$$

Note:- $y'(x_i) = g(x_i, y_i)$ can be readily computed.

Last time I had given an example to calculate P(x)



Algorithm for computing P(x)

Set
$$z_0 = x_0$$
, and $z_1 = x_0$

$$z_2 = x_1$$
, and $z_3 = x_1$

$$\vdots$$

$$z_{2i} = x_i$$
, and $z_{2i+1} = x_i$

$$\vdots$$

$$z_{2n} = x_n$$
, and $z_{2n+1} = x_n$
Set $f[z_i, z_{i+1}] = \begin{cases} f'(z_i) & \text{if } z_i = z_{i+1} \\ \frac{f(z_{i+1}) - f(z_i)}{z_{i+1} - z_i} & \text{if } z_i \neq z_{i+1} \end{cases}$

For higher order computation

$$f[z_{i}, z_{i+1}, \cdots, z_{i+k}] = \frac{f[z_{i+1}, \cdots, z_{i+k}] - f[z_{i}, \cdots, z_{i+k-1}]}{z_{i+k} - z_{i}}$$

$$P_{2n+1}(x) = f[z_{0}] + f[z_{0}, z_{1}](x - z_{0}) + f[z_{0}, z_{1}, z_{2}](x - z_{0})(x - z_{1})$$

$$+ \cdots + f[z_{0}, \cdots, z_{2n+1}] \prod_{i=1}^{2n} (x - z_{i})$$

Problem

$$f(0) = 1$$
, and $f(1) = 1.9$, $f'(0) = 2$, and $f'(1) = 2.5$
Approximate $f(0.4)$

Set
$$z_0 = 0$$
, $z_1 = 0$, $z_2 = 1$, $z_3 = 1$

Z	f(z)	f[,]	f[,,]	f[,,,]
0	1	2	-1.1	2.7
0	1	0.9	1.6	
1	1.9	2.5		
1	1.9			

Here

$$f[0,0] = f'(0) = 2$$

$$f[1,1] = f'(1) = 2.5$$

$$f[0,1] = \frac{f(1) - f(0)}{1 - 0} = \frac{1.9 - 1}{1} = 0.9$$

$$f[0,0,1] = \frac{f[0,1] - f[0,0]}{1 - 0} = 0.9 - 2 = -1.1$$

$$f[0,1,1] = \frac{f[1,1] - f[0,1]}{1 - 0} = 2.5 - 0.9 = 1.6$$

finally

$$f[0,0,1,1] = \frac{f[0,1,1] - f[0,0,1]}{1-0} = 1.6 - (-1.1) = 2.7$$

$$P_3(x) = 1 + 2(x-0) + (-1.1)(x-0)^2 + 2.7(x-0)^2(x-1)$$

$$= 1 + 2x - 1.1x^2 + 2.7x^2(x-1)$$

$$f(0.4) \approx P_3(0.4) = 1.365$$

Theory of Osculatory Interpolation

Convention:-

Let x_0, x_1, \dots, x_m be not necessarily distinct points.

We say two function f(x) and g(x) agree at the points x_0, x_1, \dots, x_m if

$$f^{(j)}(z) = g^{(j)}(z)$$
 , for $j = 0, 1, \cdots, k-1$

for every points z which occurs k times in the sequence x_0, x_1, \dots, x_m .

Example:-

$$x = 1, 2, 2, 1, 3, 2$$

f(x) and g(x) agree at the points 1, 2, 2, 1, 3, 2. If

$$f(1) = g(1)$$

$$f'(1) = g'(1)$$

$$f(2)=g(2)$$

$$f'(2)=g'(2)$$

$$f''(2) = g''(2)$$

$$f(3) = g(3)$$

Problem:-

Given x_0, x_1, \dots, x_m not necessarily distinct points and $f : [a, b] \to \mathbb{R}$.

We need a polynomial p(x) of degree $\leq m$ such that p(x) and f(x) agree at x_0, x_1, \dots, x_m .

Exercise: Show that if two polynomials of degree $\leq m$ agree at x_0, x_1, \dots, x_m , then they are equal.

So it makes sense to talk about the polynomial of degree $\leq m$ which agrees with f(x) at the m+1 points x_0, x_1, \dots, x_m .

Theorem

If f(x) has r continuous derivatives and no point in the x_0, x_1, \dots, x_m occurs more than r times then there exists exactly one polynomial $P_m(x)$ of degree $\leq m$ which agree with f(x) at x_0, x_1, \dots, x_m .

Proof

Uniqueness-Exercise

Existence:-

Assume $x_0 \le x_1 \le \cdots \le x_m$ for m = 0 there is nothing to show.

Assume the statement is true for m = k - 1 and consider it for m = k. There are two cases

Case 1 $x_0 = x_k$. Then $x_0 = x_1 = x_2 = \cdots = x_k$. So $r \ge k$. By assumption f has at least k continuous derivatives.

Then the Taylor polynomial $P_k(x)$ for f(x) around the center $c=x_0$ does the job.

Remark: Note that its leading coefficient is $\frac{f^{(k)}(x_0)}{k!}$



Continue proof

Case 2 $x_0 < x_k$

Then by induction hypothesis we can find polynomial $P_{k-1}(x)$ of degree $\leq k-1$ which agree with f(x) at x_0, x_1, \cdots, x_k and a polynomial $q_{k-1}(x)$ of degree $\leq k-1$ which agree with f(x) at x_1, x_2, \cdots, x_k .

Verify

$$P_k(x) = \frac{x - x_0}{x_k - x_0} q_{k-1}(x) + \frac{x_k - x}{x_k - x_0} P_{k-1}(x)$$

does the job.

[Slightly tricky to show. See textbook Conte and de Boor page 64.]



Convention

 x_0, x_1, \cdots, x_m not necessarily distinct points

 $P_m(x)$ = unique polynomial which agree with f(x) at x_0, x_1, \dots, x_m .

 $f[x_0, x_1, \dots, x_m] = \text{leading coefficient of } P_m(x) = \text{coefficient of } x^m \text{ in } P_m(x).$

We have

$$P_m(x) = P_{m-1}(x) + f[x_0, x_1, \cdots, x_m] \prod_{i=0}^{m-1} (x - x_i)$$

Proof:-

$$P_m(x) - f[x_0, x_1, \cdots, x_m] \prod_{i=0}^{m-1} (x - x_i)$$

has degree $\leq m-1$ and agrees with f(x) at x_0, x_1, \dots, x_m .

So by uniqueness of interpolating polynomial the result follows.



Thus we can write $P_m(x)$ as

$$P_m(x) = f(x_0) + f[x_0, x_1](x - x_0) + \cdots + f[x_0, x_1, \cdots, x_m] \prod_{i=0}^{m-1} (x - x_i)$$

Note we are assuming $x_0 \le x_1 \le \cdots \le x_m$.

Case 1
$$x_0 = x_1 = x_2 = \cdots = x_m$$
.

$$P_{m}(x) = \text{Taylor polynomial with center } x_{0}$$

$$= f(x_{0}) + f'(x_{0})(x - x_{0}) + \frac{f''(x_{0})}{2!}(x - x_{0})^{2}$$

$$+ \dots + \frac{f^{(m)}(x_{0})}{m!}(x - x_{0})^{m}$$
So
$$f[x_{0}, x_{0}, \dots, x_{0} (m + 1 \text{ times })] = \frac{f^{(m)}(x_{0})}{m!}$$

Case 2 $x_m \neq x_0$. Then

$$f[x_0, x_1, x_2, \cdots, x_m] = \frac{f[x_1, x_2, \cdots, x_m] - f[x_0, x_1, \cdots, x_{m-1}]}{x_m - x_0}$$

Example: f(0) = 1, f'(0) = 0, f''(0) = 1, f(0.1) = 9.95E - 1. Approximate f(0.05).

Z	f(z)	f[,]	f[,,]	f[,,,]
0	1	0	0.5	-10
0	1	0	-5E-1	
0	1	-5E-2		
0.1	9.95 <i>E</i> -1			

$$P(x) = 1 + 0(x - 0) + (0.5)(x - 0)^{2} + (-10)(x - 0)^{3}$$
$$= 1 + 0.5x^{2} - 10x^{3}$$
$$P(0.05) \approx 1 + 0.5(0.05)^{2} - 10(0.05)^{3} = 1$$



Calculations

$$z_0 = 0, z_1 = 0, z_2 = 0, z_3 = 0.1$$

$$f[z_0, z_1] = f'(0) = 0$$

$$f[z_1, z_2] = f'(0) = 0$$

$$f[z_2, z_3] = \frac{f(z_3) - f(z_2)}{z_3 - z_2} = -5E - 2$$

$$f[z_0, z_1, z_2] = \frac{f''(z_0)}{2} = 0.5$$

$$f[z_1, z_2, z_3] = \frac{f[z_2, z_3] - f[z_1, z_2]}{z_3 - z_1} = \frac{-5E - 2 - 0}{0.1} = -5E - 1$$

$$f[z_0, z_1, z_2, z_3] = \frac{f[z_1, z_2, z_3] - f[z_0, z_1, z_2]}{z_3 - z_0} = -10$$



Examples where Osculatory interpolation is useful

1.
$$\frac{dy}{dx} = f(x, y)$$
, and $y(x_0) = y_0$

Note that numerical methods for solving O.D.Es yield

$$\begin{array}{c|cc}
x & y \\
\hline
x_1 & y(x_1) \\
x_2 & y(x_2) \\
\vdots & \vdots \\
x_n & y(x_n)
\end{array}$$

Note that $y'(x_i) = f(x_i, y_i)$ can be calculated easily.

2.
$$f(x) = \int_0^x g(t) dt$$

Numerical integration techniques yield $f(x_0)$, $f(x_1)$, \cdots , $f(x_n)$. Note f'(x) = g(x), $f'(x_i) = g(x_i)$ can be calculated easily.



As the following example will be used often I give a direct proof.

Example: a, b distinct points. We know f(a), f(b), f'(a), f'(b).

$$P_3(x) = f(a) + f[a, a](x - a) + f[a, a, b](x - a)^2 + f[a, a, b, b](x - a)^2(x - b)$$

We prove by direct computation that $P_3(x)$ agree with f(x) at a, a, b, b

$$P_{3}(a) = f(a)$$

$$P'_{3}(a) = f[a, a] = f'(a)$$

$$f[a, b] = \frac{f(b) - f(a)}{b - a}$$

$$f[a, a, b] = \frac{f[a, b] - f[a, a]}{b - a}$$

$$= \frac{\frac{f(b) - f(a)}{b - a} - f'(a)}{b - a}$$

$$= \frac{f(b) - f(a) - (b - a)f'(a)}{(b - a)^{2}}$$

$$P_3(b) = f(a) + f'(a)(b-a) + \frac{f(b) - f(a) - (b-a)f'(a)}{(b-a)^2}(b-a)^2$$

= $f(b)$

Verify

$$f[a, b, b] = \frac{(b-a)f'(b) - (f(b) - f(a))}{(b-a)^2}$$

$$f[a, a, b, b] = \frac{(b-a)(f'(b) + f'(a)) - 2(f(b) - f(a))}{(b-a)^3}$$

$$P'_3(b) = f'(a) + 2f[a, a, b](b-a) + f[a, a, b, b](b-a)^2$$

$$= f'(b) \text{ (Check ?)}$$

Continuity of divided differences

Theorem

 $f[x_0, x_1, \dots, x_n]$ is a continuous function of x_0, x_1, \dots, x_n . (Assume f has n continuous derivatives)

i.e. for each r, $x_0^{(r)}, \dots, x_n^{(r)}$ are n+1 points in [a,b] and

$$\lim_{r\to\infty} x_i^{(r)} = y_i \text{ for } i = 0, 1, \cdots, n$$

Then

$$\lim_{r\to\infty} f[x_0^{(r)}, x_1^{(r)}, \cdots, x_n^{(r)}] = f[y_0, y_1, \cdots, y_n]$$

Proof.

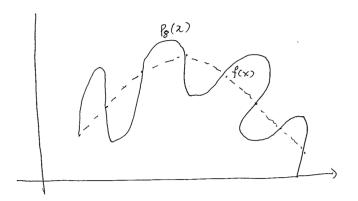
See textbook Conte and de Boor page 65.



Disadvantages of Interpolation

Note that if x_0, x_1, \dots, x_k are points in [a, b] then the interpolating polynomial has degree k.

In practice k is large. Furthermore a polynomial of degree k with k large Oscillates a lot



For example if there are 101 points then it is not advisable to work with a degree 100 interpolating polynomial as this also creates lot of round-off error.

Strategy:-

Use piecewise-polynomial approximation

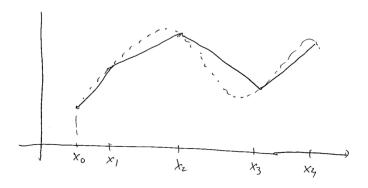
Simplest case Piecewise-liner interpolation:-

$$a = x_1 < x_2 < x_3 < \cdots < x_n = b$$

 $f(\overline{x})$ is approximated at a point \overline{x} by first locating the interval $[x_k, x_{k+1}]$ containing \overline{x} and then taking

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

Graphical representation



Note that if N is large and so $|x_i - x_{i+1}|$ is small then this is a good approximation of f(x).

If we use higher degree (say cubic) piecewise-polynomial approximation then we get better approximation.

Construction of piecewise-cubic function at the points x_1, \dots, x_{N+1} where

$$a = x_1 < x_2 < x_3 < \cdots < x_{N+1} = b$$

On each $[x_i, x_{i+1}]$ we construct $g_3(x)$ as a cubic polynomial $P_i(x)$, $i = 1, 2, \dots, N$

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

for $i = 1, 2, \dots, N$. Since $g_3(x_i) = f(x_i)$, for $i = 1, 2, \dots, N + 1$ We have

$$P_i(x_i) = f(x_i)$$
 and $P_i(x_{i+1}) = f(x_{i+1})$

for
$$i = 1, 2, \dots, N$$



In particular

$$P_{i-1}(x_i) = P_i(x_i) = f(x_i)$$
, for $i = 1, 2, \dots, N$

So $g_3(x)$ is continuous on [a, b].

Only constraints for $P_i(x)$ is

$$P_i(x_i) = f(x_i)$$
 and $P_i(x_{i+1}) = f(x_{i+1})$

So we have some freedom in choosing $P_i(x)$

We study 2 cases

- Piecewise-Cubic Hermite interpolation
- Cubic-Spline interpolation



Piecewise-Cubic Hermite interpolation

One determine $P_i(x)$ so as to interpolate f(x) at $x_i, x_i, x_{i+1}, x_{i+1}$, *i.e.* we also have

$$P'_i(x_i) = f'(x_i)$$
 and $P'_i(x_{i+1}) = f'(x_{i+1})$

By Newtone formula

$$P_{i}(x) = f(x_{i}) + f[x_{i}, x_{i}](x - x_{i}) + f[x_{i}, x_{i}, x_{i+1}](x - x_{i})^{2}$$

$$+ f[x_{i}, x_{i}, x_{i+1}, x_{i+1}](x - x_{i})^{2}(x - x_{i+1})$$
Write $x - x_{i+1} = (x - x_{i}) + (x_{i} - x_{i+1})$

$$P_{i}(x) = f(x_{i}) + f'(x_{i})(x - x_{i})$$

$$+ (f[x_{i}, x_{i}, x_{i+1}] - f[x_{i}, x_{i}, x_{i+1}, x_{i+1}] \triangle x_{i})(x - x_{i})^{2}$$

$$+ f[x_{i}, x_{i}, x_{i+1}, x_{i+1}](x - x_{i})^{3}$$

Algorithim

For
$$i = 1, 2, \dots, N + 1$$

$$f_{i} = f x_{i}$$

$$\triangle x_{i} = x_{i+1} - x_{i}$$

$$s_{i} = f'(x_{i}), c_{1,i} = f_{i}, c_{2,i} = s_{i}$$

$$c_{3,i} = f[x_{i}, x_{i}, x_{i+1}] - f[x_{i}, x_{i}, x_{i+1}, x_{i+1}] \triangle x_{i}$$

$$= \frac{f[x_{i}, x_{i+1}] - s_{i}}{\triangle x_{i}} - c_{4,i} \triangle x_{i}$$

$$c_{4,i} = f[x_{i}, x_{i}, x_{i+1}, x_{i+1}]$$

$$= \frac{f[x_{i}, x_{i+1}, x_{i+1}] - f[x_{i}, x_{i}, x_{i+1}]}{\triangle x_{i}}$$

$$= \frac{s_{i+1} + s_{i} - 2f[x_{i}, x_{i+1}]}{\triangle x_{i}}$$

Example

$$\frac{dy}{dx} = y - x^2 + 1$$
, and $y(0) = 0.5, 0 \le x \le 1$

x_i	$y(x_i)$	$y'(x_i)$
0	0.5	1.5
0.2	0.826	1.786
0.4	1.207	2.047
0.6	1.637	2.277
8.0	2.110	2.470
1.0	2.618	2.618

Find y(0.7), y(0.9)

Note:- $y(x_i)$ is found using a Numerical method

Remarks:- Usual oscillatory polynomial has degree 11.



So we use piecewise Hermite interpolation

X	y(x)	f[,]	f[,,]	f[,,,]
0.6	1.637	2.277	4.4 <i>E</i> -1	4.25 <i>E</i> −1
0.6	1.637	2.365	5.25 <i>E</i> -1	
8.0	2.110	2.470		
0.8	2.110			

$$P_4(x) = 1.637 + 2.277(x - 0.6) + 4.4E - 1(x - 0.6)^2 + 4.25E - 1(x - 0.6)^2(x - 0.8)$$

$$P_4(0.7) = 1.869$$

X	<i>y</i> (<i>x</i>)	f[,]	f[,,]	f[,,,]
0.8	2.110	2.47	0.35	0.2
8.0	2.110	2.54	0.39	
1.0	2.618	2.618		
1.0	2.2.618			

$$P_4(x) = f[0.8] + f[0.8, 0.8](x - 0.8) + f[0.8, 0.8, 1](x - 0.8)^2 + f[0.8, 0.8, 1, 1](x - 0.8)^2(x - 1) = 2.11 + 2.47(x - 0.8) + 0.35(x - 0.8)^2 + 0.2(x - 0.8)^2(x - 1) P_5(0.9) = 2.360 in 4 sig digits$$