

Engineering Mathematics II
Professor Jitendra Kumar
Department of Mathematics
Indian Institute of Technology, Kharapur
Lecture 26
Polynomial Interpolation

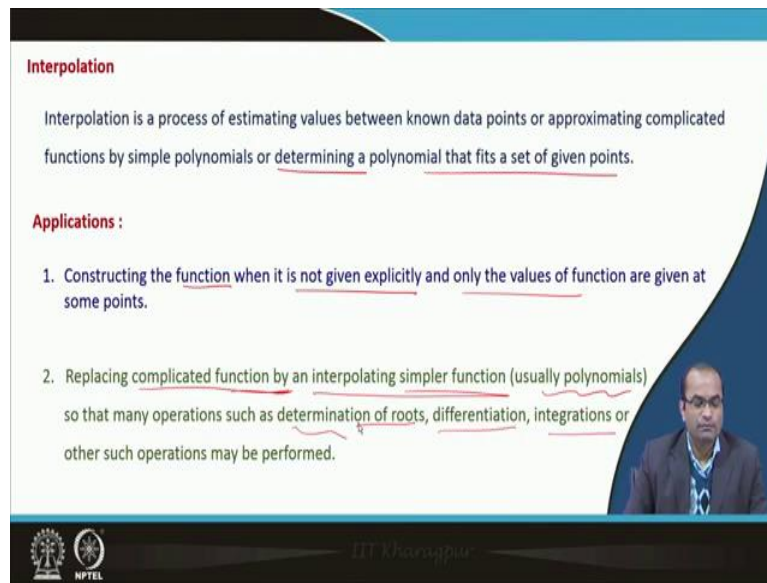
So, welcome back to lectures on Engineering mathematics 2 and this is lecture number 26 on polynomial interpolation. A very important topic in numerical analysis.

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And in this lecture we will be talking about what is actually the polynomial interpolation and then we will be also talking about the existence and the uniqueness of such polynomial interpolation and finally we will derive a formula which will provide us the error in such interpolating polynomial.

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Interpolation

Interpolation is a process of estimating values between known data points or approximating complicated functions by simple polynomials or determining a polynomial that fits a set of given points.

Applications :

1. Constructing the function when it is not given explicitly and only the values of function are given at some points.
2. Replacing complicated function by an interpolating simpler function (usually polynomials) so that many operations such as determination of roots, differentiation, integrations or other such operations may be performed.

The slide includes a video inset of a man in a suit and glasses speaking. At the bottom left are the logos for IIT Kharagpur and NPTEL. The name 'Dr. Karan Singh' is visible at the bottom center.

So, coming to the interpolation, it is a process of estimating values between known data points or approximating complicated functions by simple polynomials or determining a polynomial that fits a set of given data points.

So, it is a various ways one can define the interpolation, so the applications they are like constructing the function when it is not given explicitly only the values of functions are given, so we will construct a polynomial which passed through or fit the given data point, so that is the one of the major applications of these interpolation.

The second one that replacing the complicated functions were by an interpolation simple functions usually the simple functions we take polynomials because polynomials are really simple function because many operations on these polynomials such as the determination of roots, differentiation, integration, and many other operations can be performed easily.


So, given a complicated function we will approximate it by some polynomial of some degree and then we can approximate basically the roots or we can do differentiation, integration et cetera such operations can be performed.

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Fundamental principle behind polynomial interpolation

Weirstrass Approximation Theorem

Suppose that f is defined and continuous on $[a, b]$. For each $\epsilon > 0$, there exists a polynomial $P(x)$ with the property that


$$|f(x) - P(x)| < \epsilon; \forall x \in [a, b].$$


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So, what is the fundamental principle behind this polynomial interpolation that will be clear from this Weirstrass approximation theorem, which says that suppose this f is defined and it is continuous on some interval then for each epsilon however small, for each epsilon there exist a polynomial $P(x)$ with the property that $f(x) - P(x)$ is less than epsilon.

So, this relation tells that given $f(x)$ here a continuous function on a, b we can have a polynomial which will approximate this function which as good accuracy as we want because epsilon we can fix and it can be very very small and then we can find a polynomial $P(x)$ corresponding to this given continuous function $f(x)$.

So, that the base of this interpolation or having a polynomial approximation of a given function. Because this theorem says that for any given function which is continuous we can find a polynomial with as good accuracy as we want. And based on these Weirstrass approximation theorem then we will derive the polynomials corresponding to a given function.

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Why not Taylor's Polynomial ? Consider Taylor's Polynomial of e^x around $x = 0$.

$P_0(x) = 1;$

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$P_5(x) = 1 + x + \frac{x^2}{2} + \frac{x^3}{6} + \frac{x^4}{24} + \frac{x^5}{120}$

$P_4(x) = 1 + x + \frac{x^2}{2} + \frac{x^3}{6} + \frac{x^4}{24}$

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Taylor's polynomials agree as closely as possible with a given function at a specific point, so they concentrate their accuracy near that point.

For ordinary computation purposes it is more efficient to use methods that include information at various points.

Dr. Shivaprasad

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So, but the question is that but why not Taylor's polynomial, because we have already studied Taylor, Taylor's polynomial which was also used for approximating a given function. And now we are talking about some other polynomial with not exactly the Taylor's polynomial so the question is that, why not just the Taylor's approximation or the Taylor's polynomial because they are also doing the same job given a function we are approximating a function in terms of the polynomial which is in this case Taylor's polynomial.

So, let us just demonstrate through this example and then we will give the reason that why not Taylor's polynomial we are talking about some other kind of polynomials here. So, constructing the Taylor's polynomial for instance this e^x the exponential function around this x equal to 0. So, the first polynomial of this degree 0 that is a constant function that is 1.

Because that is just function value there so $f(0) = e^0 = 1$. And if we plot this so this was the exponential function here and this is this constant function so it is having very good accuracy at this point or naturally it is matching with the function itself at x is equal to 0. Other than that point the accuracy is not as good and it is actually getting worse when we are going far away from this point. So, the accuracy is more or less concentrated around this point of expansion that is the one point here we should not.

If we go ahead we construct this polynomial of degree 1 so that would be $1 + x$ that is the Taylor's polynomial here. And if we plot this now it is linear function $1 + x$ and that is given here. And now we have higher degree polynomial and then we are getting better

approximation in the neighborhood again of that point but if we go far again this approximation is not at all good.

If we construct further, for example, second order polynomial or the third order polynomial so we plot here the third order polynomial for the instance now we are matching quite a bit in in a larger neighborhood around this point x equal to 0 and similarly, if continue and construct for instances fifth order polynomial then we do c here the accuracy is now in a much wider interval.

So, what is the point now here? I mean the question is that why not this Taylor's polynomial, we have that Taylor's polynomial here which is approximating the given function for instance the exponential function. So, the main issues here the Taylor's polynomial agrees as closely as possible with a given function at a specific point which was the point of expansion or basically in other word we say that they concentrate their accuracy near that point. So, as we are going far away from that point the accuracy is not at all good but in the close neighborhood of that point the accuracy is very good.

The second point here to construct the Taylor's polynomial we do need information of that function at x is equal to 0. So, we need the function value at that point, we need its derivative, second derivative, third derivative and so on. Depending on what degree polynomial we want to construct.


Whereas in the interpolation we will again, we will be constructing the polynomials but there we need the function value at different different points not at one point the values of higher order derivatives rather we need the function value at different different points, so we will naturally come up with some other polynomial not the Taylor's polynomial which also does the purpose of approximating a complicated function by a polynomial.



So, for ordinary computation purposes, it is more efficient to use methods that include information at various points not at just particular point because in many applications, many experimental results we have at different different points the given data and then we want to know that which polynomial fits the best, the given data points.

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Existence and uniqueness for polynomial interpolation

For $(n + 1)$ data points there is one and only one polynomial of order $\leq n$ that passes through all the points.







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Existence and uniqueness for polynomial interpolation

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For example, there is only one straight line (a first order polynomial) that passes through two points.



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Existence and uniqueness for polynomial interpolation


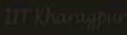

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For example, there is only one straight line (a first order polynomial) that passes through two points.

Consider for simplicity, a second order polynomial

$$f(x) = a_0 + a_1x + a_2x^2$$

A straight forward method for computing the coefficients of a polynomial of degree n is based on the fact that $(n + 1)$ data points are required (3 data points in this example) to determine $(n + 1)$ unknowns (3 unknowns a_0, a_1, a_2 in this example)

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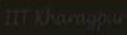

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So, the first question is whether given data points the polynomial which we want to fit is it unique and does it always possible to construct such a polynomial. So, for n plus 1 data points there is one and only one polynomial of degree less than or equal to n that passes through all the points. That what we are going to prove here in this result so that is the main result that there it exist and it is unique so when n plus 1 data points are given there is only one and there is one only one polynomial of degree this less than or equal to n that passes through all the points.

A very simple example, which we can talk about that when we are talking about the two points and we know that, there is unique straight line. So, a polynomial of degree 1 which passes through these given two points they cannot be two lines, two different lines which passes through both the points. So, it is a uniqueness and there will always a line so there is

existence so there is only one straight line a first order polynomial in this case when two data points are given that passes through the given two points.

So, now to prove this result which is true for general n, we will consider a second order polynomial we will prove this for a second order polynomial and one can extend this for nth degree or order polynomial. So, suppose our function here with the polynomial $a_0 + a_1x + a_2x^2$ a second order polynomial we have here.

So, how many unknowns do we have? We have a_0 unknown, a_1 and a_2 . So, there are three, there are three unknowns which will define this polynomial so to, in order to compute these three unknowns we basically need three data points. The point is a straight forward method for computing the coefficients of the polynomial of degree n.

So, if we talk in general n based on the fact n plus 1 data points are required like in this case it is second degree polynomial and we need three data points because there are three unknowns and for each data points we can construct one linear equation in a_0, a_1, a_2 so we need three data points to have three equations and they can be solved and we will see in this particular case that there exist a unique solution. So, for this uniqueness we need n plus 1 data points and three data points in this example to determine n plus 1 unknowns and in this example we have three unknowns for instance.

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Polynomial to be fitted with the given data $f(x) = a_0 + a_1x + a_2x^2$

Suppose that there are 3 given data points $(x_0, f(x_0)), (x_1, f(x_1)), (x_2, f(x_2))$.

If the given polynomial passes through the given data points then it must satisfy them, i.e.,

$f(x_0) = a_0 + a_1x_0 + a_2x_0^2$

$f(x_1) = a_0 + a_1x_1 + a_2x_1^2$

$f(x_2) = a_0 + a_1x_2 + a_2x_2^2$

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Polynomial to be fitted with the given data $f(x) = a_0 + a_1x + a_2x^2$

Suppose that there are 3 given data points $(x_0, f(x_0)), (x_1, f(x_1)), (x_2, f(x_2))$.

If the given polynomial passes through the given data points then it must satisfy them, i.e.,

$$\begin{cases} f(x_0) = a_0 + a_1x_0 + a_2x_0^2 \\ f(x_1) = a_0 + a_1x_1 + a_2x_1^2 \\ f(x_2) = a_0 + a_1x_2 + a_2x_2^2 \end{cases} \Leftrightarrow \begin{bmatrix} 1 & x_0 & x_0^2 \\ 1 & x_1 & x_1^2 \\ 1 & x_2 & x_2^2 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} f(x_0) \\ f(x_1) \\ f(x_2) \end{bmatrix}$$

So, now this is the polynomial to be fitted with given data and suppose there are the 3 data points $x_0, f(x_0), x_1, f(x_1),$ and $x_2, f(x_2)$, so these 3 data points are given and we want to construct a polynomial which actually passes through these 3 points. So, if the given polynomial passes through the given data points then it must satisfy that means the $f(x_0)$ if we pass here x_0 point so $f(x_0)$ must be equal to a_0 plus $a_1 x_0$, plus $a_2 x_0$ square.

So, that is the first condition we should have that these polynomial should satisfy the given data point. The second data point is $x_1, f(x_1)$, so when we substitute in the polynomial x_1 for x then we should get $f(x_1)$ and the third one when we substitute this x_2 we should get $f(x_2)$. So, these are three equations we have from this given polynomial which passes through the three given points.

This we can also rewrite in the matrix vector form so here the matrix is then the coefficient of a_0 that is 1 here, the coefficient of a_1 that is x_0 , the coefficient of a_2 that is x_0 square. Similarly, here we have the coefficient of this a_0 that is 1, so it is not 2 it is 1 and then we have x_1 there, we have x_1 square and in this case again we have 1, we have x_2 and we have x_2 square. So, this is the system we have for corresponding to this given linear equations.

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$$\begin{pmatrix} 1 & x_0 & x_0^2 \\ x_1 & x_1^2 \\ x_2 & x_2^2 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \end{pmatrix} = \begin{pmatrix} f(x_0) \\ f(x_1) \\ f(x_2) \end{pmatrix}$$

This system of equations has a unique solution as

$$\text{Det} \begin{pmatrix} 1 & x_0 & x_0^2 \\ x_1 & x_1^2 \\ x_2 & x_2^2 \end{pmatrix} = (x_0 - x_1)(x_1 - x_2)(x_2 - x_0) \neq 0 \text{ if } x_0, x_1, x_2 \text{ are distinct.}$$

In practice, it is observed that the above system of equations is ill-conditioned.

Whether they are solved with an elimination method or with a more efficient algorithm, the resulting coefficient can be highly inaccurate, in particular for large n .

Therefore, we have some mathematical formats (interpolating formats) in which such calculation can be avoided.

So, having this system now, what we will analyze now that this system of equations has a unique solution, this we will show now in a minute. And when we compute the determinant of this function here what we will realize that the value of the determinant is nothing but x_0 minus x_1 , x_1 minus x_2 and x_2 minus x_0 .

So, having this value what we realize that if these points x_0 , x_1 , x_2 are different, if these points are different, what will happen that the value here will be non-zero so the value of this will be non-zero and once we have this determinant non-zero we know that there will be a unique solution.

So, if these are distinct and they are obviously distinct we are talking about x_0 , x_1 , x_2 they are distinct points then definitely we have this unique solution. So, in practice. What is the problem now? So, this is what we can, this is, this explains the method also the way we can construct a polynomial given data points.

So, if $n + 1$ data points are given we can construct uniquely a polynomial of degree n . So, what is the problem why we are now discussing more about the polynomial, the reason is that in practice it is observed that the above system of equations is ill conditioned. What do we mean by ill conditioned here?

So, when, whether they are solved, so this system or the equation they are solved with an elimination method Gauss elimination et cetera or more they like iterative methods we have

already discussed. So, any other efficient algorithm we take the resulting coefficient can be highly inaccurate.

So, this ill conditioned means very little error in the, in this matrix entries can lead to a very inaccurate, highly inaccurate values of its solution that means a naught, a1 and a2 so that means the ill conditioned so this is what observed we are not going much into detail of this part.

But for large n, one can understand that the system is actually difficult to solve because of this ill conditioned system. So, what we have then there are some mathematical formats which we will be talking about in the next lecture and those are the methods how to get this interpolating polynomial without following this approach that constructing the system of linear equations and then solving them. Rather we have a better ways and direct ways of constructing the polynomial without solving such a system of linear equations.

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Error in Interpolating Polynomials

Let x_0, x_1, \dots, x_n be $(n+1)$ points and let x be a point belonging to the domain of a given function f .

Assume that $f \in C^{(n+1)}(I_x)$, where I_x is the smallest interval containing the nodes x_0, x_1, \dots, x_n and x .

Then the interpolation error at the point x is given by

$$E_n(x) = f(x) - P_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} (x-x_0)(x-x_1)\cdots(x-x_n) \text{ where } \xi \in I_x.$$

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  Dr. Khosla

Well, so the another, the last portion of this lecture we will be talking about the error in the interpolating polynomials. Because the polynomial is unique and now we can talk about before we go for different methods how to construct that polynomial because the polynomial is unique so we can talk about what is actually the error in this approximation that is also an important point which we should address.

So, suppose x naught, x_1, x_2, x_3, x_n these are n plus 1 points. And we let that x be a point belonging to the domain of a given function. So, all these points x_0, x_1, x_2 et cetera and also we have taken one more point x these all belongs to domain of the function. And we assume that f is n plus 1 times differentiable function where this I_x is the domain here or the interval containing all these notes and this x is a smallest interval we can think of which contains all these points including this x . So, I_x is basically the domain of these function f .


So, then the interpolation error at the point x , so because we are talking about the polynomial which or the function which has its domain. So, at any point x difference between the polynomial and the function is given by so $f(x)$ minus this $P_n(x)$ one can compute by this formula here which has this n plus 1th order derivative so f the n plus 1th order derivative at point ξ which belong to this interval itself factorial n plus 1. And there is a product here x minus x naught, x minus x_1, x minus x_n . So, this is the error we have when we interpolate this function, approximate this function by the polynomial taking these n plus 1 points.


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Error in Interpolating Polynomials

$$E_n(x) = f(x) - P_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} (x-x_0)(x-x_1)\cdots(x-x_n) \text{ where } \xi \in I_x.$$

Proof: Note that the result is obviously true if x coincides with any of the interpolating nodes.




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Error in Interpolating Polynomials

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For simplicity, let us assume $w_{n+1}(x) = (x-x_0)(x-x_1)\cdots(x-x_n)$



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Error in Interpolating Polynomials

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Proof: Note that the result is obviously true if x coincides with any of the interpolating nodes.

For simplicity, let us assume $w_{n+1}(x) = (x-x_0)(x-x_1)\cdots(x-x_n)$

Now, define for any $x \in I_x$, the function

$$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$$



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Error in Interpolating Polynomials

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$$E_n(\xi) = f(\xi) - P_n(\xi)$$

Since $f \in C^{(n+1)}(I_x)$ and w_{n+1} is a polynomial, then $G \in C^{(n+1)}(I_x)$.



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Error in Interpolating Polynomials

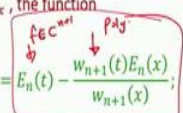
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So, we want to derive now how to get this formula here for the error so the proof we will just go for the outline, so the result is obviously true when x coincides with the any of the interpolating nodes. Why? So, if we are talking about that this x coincides with x_i then we have here x_i and this also x_i and naturally this polynomial passes through all these points $f(x_i)$ so this will be 0 here, so then the error is 0 and then here this side also this formula will give because when x is one of this nodes so this will make everything 0 so we have 0 equal 0.

So, that is obviously true when this x coincide with the interpolating nodes meaning this x naught, x_1, x_2, x_3 and so on. And now for simplicity we assume because there was a product here lengthy product x minus x naught, x minus x_1, x minus x_n . So, this product here we are denoting with the W_{n+1} , $n+1$ is actually the degree of this polynomial again because this product is nothing but a polynomial and that degree of this polynomial is $n+1$ so we are we are denoting this by W_{n+1} .

And now we define that for any x we take in the domain of this function where we want to calculate this difference between the two, we define this function and this function will to help us prove exactly this formula for the error which is the aim this lecture now. So, we define this function G_t as $E_n(t)$, E_n is defined already that is the difference between this $f(x)$ and the polynomial $P_n(x)$.

So, here we have $E_n(t)$ then use here this $W_{n+1}(t)$ which is already define here as $w_n(x)$, we have again E_n and we have again W_{n+1} so in terms of this E_n and W_{n+1} which is also W_{n+1} is defined here and E_n is defined there, so in terms of this W_{n+1} , E_n we have defined a function G_t . And now we will discuss what are the properties of this G_t and how we can get out of with the help of this function the error formula.

So, the first thing we should notice that since this f is $n+1$ times differentiable function this G is also $n+1$ times differentiable function and why? Because G_t is nothing but the $E_n(t)$, $E_n(t)$ and $W_{n+1}(t)$, $E_n(x)$ and this $W_{n+1}(x)$ they are constants because for any but fixed x here we are talking about this formula here t is only the variable which we are considering now.

So, what is $E_n(t)$? $E_n(t)$ is $f(t)$ minus this $P_n(t)$. So, again this is a polynomial and f is a $n+1$ th time differentiable function so E_n we have here the $n+1$ time differentiable so there is no problem for the differentiability this E_n for, sorry for this G_t as long as this first term is

concerned and the second one here this is polynomial these are some fixed values so this is fixed for given x.

And w_n is a polynomial so here we have this polynomial, here we have that function which is again $n + 1$ time differentiable so everything here we can differentiate $n + 1$ times that means the G function is $n + 1$ times differentiable. That is the first property of this function G .

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$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$

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

$w_{n+1}(x) = (x - x_0)(x - x_1) \cdots (x - x_n)$

Note that $G(t)$ has $(n + 2)$ distinct zeros in I_x since

$G(x_i) = E_n(x_i) - \frac{w_{n+1}(x_i)E_n(x)}{w_{n+1}(x)} = 0$

$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$

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

$w_{n+1}(x) = (x - x_0)(x - x_1) \cdots (x - x_n)$

Note that $G(t)$ has $(n + 2)$ distinct zeros in I_x since

$G(x_i) = E_n(x_i) - \frac{w_{n+1}(x_i)E_n(x)}{w_{n+1}(x)} = 0; i = 0, 1, 2, \dots, n.$

$G(x) = E_n(x) - \frac{w_{n+1}(x)E_n(x)}{w_{n+1}(x)}$

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$$G(x) = E_n(x) - \frac{w_{n+1}(x)E_n(x)}{w_{n+1}(x)} = 0$$

Then using Rolle's theorem, G' has at least $(n + 1)$ distinct zeros.

By recursion it follows that $G^{(j)}$ admits at least $(n + 2) - j$ distinct zeros.

$\Rightarrow G^{(n+1)}$ has at least one zero, which we denote by ξ , i.e., $G^{(n+1)}(\xi) = 0$

So, having this G we have En we have Wn already and what we will notice now that Gt this function has n plus 2 distinct zeros in this Ix. How these, how this function has n plus 2 distinct zeros we will notice now, so, just consider that G xi so we have En xi we have here also Wn plus 1 xi. Now, this Wn plus 1 xi this is a Wn plus 1, when we put here for x equal to xi, this will become 0, so this part is 0, and hence this part is 0 so the second part is 0.

Now, coming to the first one, when En xi this error here at xi naturally that will be 0 because this xi is the narrow point. So, this is also 0 and 0 minus 0 then everything is 0 here for all i is 0, 1, 2, 3 and so on n. So, there are here n plus 1 points, so here we have n plus 1 points where this G xi vanishes, G vanishes so there are n plus 1 distinct points we have discuss and one more point that is the x which we have already fixed at the beginning its arbitrarily point but fixed point here so this at Gx also notice we will notice that this is equal to 0 because we have now En x, Wn and so on, these two are the same so cancel out we have En x minus En x and that is 0.

So, here the Gx is also 0, so we have n plus 1 points here, n plus 1, n plus 1 so the total there are n plus 2 distinct zeros in I x of this G. So, having this property now, we can use the Rolle's theorem which will say that G prime its derivative will have at least n plus 1 distinct points because whenever we have a function for instance it has these two zeros here so if Fa is 0 and Fb is 0, then the Rolle's theorem says that there will be a point in between where its derivative will be also 0.

So, between the two zeros here, its derivative will have one zeros and the same argument we can have here so the G function G is 0 at x0 and x1, x2 and so on and also xn and also there is

a point x somewhere. So, there are $n + 2$ points there and between each these two zeros there will be 1 one point where G prime will be 0 and so on and we count this will be just a 1 less than the number of these zeros.

So, there were $n + 2$ zeros, so G prime will have $n + 2$ minus 1 that is $n + 1$ distinct zeros. By recursion the we can continue this process that G double prime will have n distinct zeros and so on or i th derivative of this G this is i th derivative of G it will admit $n + 2$ and minus the here if it is j th so let us say j th derivative then $n + 2$ minus j th distinct zeros this will have.

So, $n + 1$ now we are continuing and now let us say if j is $n + 1$ so if we put here $n + 1$ so this will remain just 1 there, so $G^{(n+1)}$ will have at least zeros this is the implication of this Rolle's theorem now and which denotes and so we are denoting by this ξ this $n + 1$ the $G^{(n+1)}$ th derivative has one zero and that is we are consider that is ξ that means that $G^{(n+1)}(\xi) = 0$ so this result we have.

(Refer Slide Time: 26:29)

The slide contains the following mathematical content:

- Equation 1: $G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$
- Equation 2: $w_{n+1}(x) = (x - x_0)(x - x_1) \cdots (x - x_n)$
- Equation 3: $G^{(n+1)}(\xi) = 0$
- Equation 4: $\Rightarrow G^{(n+1)}(t) = E_n^{(n+1)}(t) - \frac{w_{n+1}^{(n+1)}(t)E_n(x)}{w_{n+1}(x)}$
- Note: Note that $E_n(t) = f(t) - P_n(t) \Rightarrow$

In the bottom right corner, there is a video feed of a man in a suit and glasses, presumably the instructor.

At the bottom left, there are logos for IIT Kharagpur and NPTEL. At the bottom center, the name "Dr. Khosla" is visible.

$$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$$

$$w_{n+1}(x) = (x-x_0)(x-x_1)\dots(x-x_n)$$

$$G^{(n+1)}(\xi) = 0$$

$$\Rightarrow G^{(n+1)}(t) = E_n^{(n+1)}(t) - \frac{w_{n+1}^{(n+1)}(t)E_n(x)}{w_{n+1}(x)}$$

$n+1 \rightarrow \dots$
 $(n+1)!$

Note that $E_n(t) = f(t) - P_n(t) \Rightarrow E_n^{(n+1)}(t) = f^{(n+1)}(t)$ as $P_n^{(n+1)}(t) = 0$

$$w_{n+1}^{(n+1)}(t) = (n+1)!$$



$$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$$

$$w_{n+1}(x) = (x-x_0)(x-x_1)\dots(x-x_n)$$

$$G^{(n+1)}(\xi) = 0$$

$$\Rightarrow G^{(n+1)}(t) = E_n^{(n+1)}(t) - \frac{w_{n+1}^{(n+1)}(t)E_n(x)}{w_{n+1}(x)}$$

Note that $E_n(t) = f(t) - P_n(t) \Rightarrow E_n^{(n+1)}(t) = f^{(n+1)}(t)$ as $P_n^{(n+1)}(t) = 0$

$$w_{n+1}^{(n+1)}(t) = (n+1)! \quad \& \quad G^{(n+1)}(\xi) = 0$$



$$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; t \in I_x$$

$$w_{n+1}(x) = (x-x_0)(x-x_1)\dots(x-x_n)$$

$$G^{(n+1)}(\xi) = 0$$

$$\Rightarrow G^{(n+1)}(t) = E_n^{(n+1)}(t) - \frac{w_{n+1}^{(n+1)}(t)E_n(x)}{w_{n+1}(x)}$$

$t = \xi$

Note that $E_n(t) = f(t) - P_n(t) \Rightarrow E_n^{(n+1)}(t) = f^{(n+1)}(t)$ as $P_n^{(n+1)}(t) = 0$

$$w_{n+1}^{(n+1)}(t) = (n+1)! \quad \& \quad G^{(n+1)}(\xi) = 0$$

$$\Rightarrow 0 = f^{(n+1)}(\xi) - \frac{(n+1)! E_n(x)}{w_{n+1}(x)}$$



$$G(t) = E_n(t) - \frac{w_{n+1}(t)E_n(x)}{w_{n+1}(x)}; \quad t \in I_x$$

$$w_{n+1}(x) = (x - x_0)(x - x_1) \cdots (x - x_n)$$

$$G^{(n+1)}(\xi) = 0$$

$$\Rightarrow G^{(n+1)}(t) = E_n^{(n+1)}(t) - \frac{w_{n+1}^{(n+1)}(t)E_n(x)}{w_{n+1}(x)}$$

Note that $E_n(t) = f(t) - p_n(t) \Rightarrow E_n^{(n+1)}(t) = f^{(n+1)}(t)$ as $p_n^{(n+1)}(t) = 0$

$$w_{n+1}^{(n+1)}(t) = (n+1)! \quad \& \quad G^{(n+1)}(\xi) = 0$$

$$\Rightarrow 0 = f^{(n+1)}(\xi) - \frac{(n+1)!E_n(x)}{w_{n+1}(x)} \Rightarrow E_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} w_{n+1}(x)$$

So, what do we have? We $G_{n+1}(\xi) = 0$, w_{n+1} was defined by this product and we have this function which we have started with. Now, we can talk about the $n+1$ th derivative of this that is E_{n+1} th derivative and the $n+1$ th derivative of this are w_{n+1} the rest here they are just fixed values.

Now, we should note that talking about this E_n . So, $E_n(t) = f(t) - p_n(t)$ and when we take the n th order derivative so n th order derivative will go to f and this is a n degree polynomial, $n+1$ th degree, $n+1$ th derivative will be 0 of this polynomial so what we have that E_n the $n+1$ th derivative will be simply the $n+1$ th derivative of f because this is equal to 0.

What else we have the $n+1$ th derivative here of this w_{n+1} , w_{n+1} is again this is $n+1$ th degree polynomial so here we have something x^{n+1} plus anything else there with lower degree terms and when we take the $n+1$ th order derivative so simply we will get $n+1$ factorial which is written here so that is done and obviously we have this result $n+1$ th derivative of G at ξ is equal to 0.

So, having all this result now we will substitute here for t is equal to ξ so if we put if put here t is equal to ξ what will happen $G_{n+1}(\xi)$ will become 0. Then here we have this $n+1$ th derivative of E that is nothing but the $n+1$ th derivative of f so we have here $f^{(n+1)}(\xi)$ and then here also this $n+1$ factorial, then we have here E_{n+1} over w_{n+1} .

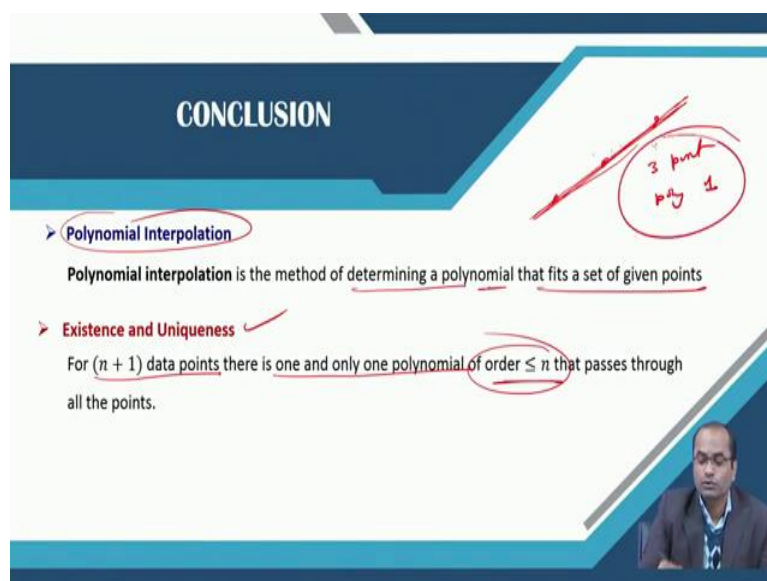
So, this is what we have now from this given function which we have chosen and now just from this we can conclude that $E_n(x)$ so this is $E_n(x)$, $E_n(x)$ will be equal to this one and this is exactly the error formula which we want to derive. So, now in this lecture we have also derived the error formula because this polynomial is unique so irrespective how do we calculate but we know that will be the error.

(Refer Slide Time: 29:09)



So, these are the references we have used for preparing this lecture


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CONCLUSION

- **Polynomial Interpolation**
Polynomial interpolation is the method of determining a polynomial that fits a set of given points
- **Existence and Uniqueness**
 For $(n + 1)$ data points there is one and only one polynomial of order $\leq n$ that passes through all the points.
- **Error in Interpolating Polynomials**

$$E_n(x) = f(x) - P_n(x) = \frac{f^{(n+1)}(\xi)}{(n+1)!} (x - x_0)(x - x_1) \dots (x - x_n)$$



And just to conclude that we introduced the polynomial interpolation which is the method of determining polynomial that fits a given set of data points. And we have also proved the existence and the uniqueness for n plus 1 data points there is only one and only one polynomial of order less than equal to n . Why less than equal n we are talking about?

Because for instance if we take, if we take three points, so according to our discussion we should have a polynomial of degree 2. But in this particular case because they are points in one line so will get only linear equation that means only the line because they were in one line co-linear points.

So, therefore, so here again having three points but we are getting a polynomial of degree 1. So, that what we have written here that is always possible that when n plus 1 data points are given there is one and only one so uniqueness is always there. And there would be polynomial of degree less than or equal to n that passes through all the points.

And we have also discussed the error in the interpolating polynomial which is given by this formula $f^{(n+1)}$ at point some X_i and then we have n plus 1th factorial and this product x minus x naught, x minus x_1 and x minus x_n . So, that is all for this lecture and I thank you very much for your attention.