MA 214 - Introduction to Numerical Analysis

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1 Interpolation Theory

1.1 Introduction

- Given finite set of points, reconstructing the original curve is interpolation.
- There will be obviously infinitely many curve.
- Interpolation problem Given n+1 real distinct points: x_0, x_1, \ldots, x_n and real numbers: y_0, y_1, \ldots, y_n Find a function $f: \mathbb{R} \to \mathbb{R}$ such that

$$f(x_i) = y_i$$
 for $i = 0, 1, ..., n$

Such a function is called **interpolant** and points x_i are called **interpolation points**. We attempt to rebuild original function using polynomial functions. This is called polynomial interpolation and function is polynomial interpolant.

• A polynomial is function of the form

$$p(x) = a_0 + a_1 x + \ldots + a_n x^n$$

• \mathbb{P}_n is the set of polynomials consisting of all polynomials of degree $\leq n$

1.2 Polynomial Interpolation

Theorem (Joseph-Louis Lagrange Theorem). Given n+1 data points with unique $x_i s$, then there exists a unique polynomial $p_n \in \mathbb{P}_n$ such that

$$p(x_i) = y_i$$
 for $i = 0, 1, ..., n$

Proof. (1) This can be shown by linear algebra. In a n degree polynomial, we substitute the points and get n+1 equations in n+1 variables (coeff) and all the rows are unique (since x_0, x_1, \ldots, x_n are unique), hence in AX = b, $|A| \neq 0$.

Proof. (2) Part 1: Uniqueness: If there is an interpolant, then the interpolant is unique Let there be 2 interpolants, p_n and q_n and let r(x) = p(x) - q(n),

$$r(x) = 0 \text{ for } i = 0, 1, \dots, n$$

This contradicts the fundamental theorem of Algebra. (A polynomial of degree n can have at most n real roots). Therefore

$$r(x) = 0 \quad \forall x \in \mathbb{R}$$

 $p(x) = q(x) \quad \forall x \in \mathbb{R}$

Part 2: Existence (construction):

Given n+1 data points, build n+1 Langrange polynomials

$$L_k^n(x_i) = \begin{cases} 0 & \text{for } i \neq k \\ 1 & \text{for } i = k \end{cases}$$

$$L_k^n(x) = \frac{(x - x_0)...(x - x_{k-1})(x - x_{k+1})...(x - x_n)}{(x_k - x_0)...(x_k - x_{k-1})(x_k - x_{k+1})...(x_k - x_n)}$$

$$p(x) = \sum_{k=0}^n y_k L_k^n(x)$$

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1.3 Closeness between functions

Given two continuous functions $f, g : [a, b] \to \mathbb{R}$, to evaluate how close the functions are consider the following

$$\max_{x \in [a,b]} |f(x) - g(x)|$$

1.4 Set of continuous Functions

C[a, b] is the set of all continuous functions on [a,b] C[a, b] is a infinite dimensional vector space

$$f, g \in C \Longrightarrow f + g \in C \text{ and } \lambda f \in C$$

We define norm on C[a, b] as

$$||f|| = \max_{x \in [a,b]} |f(x)|$$

 $C^{k}[a,b]$ denotes the set of all functions which are continuously k-times differentiable

1.5 Polynomial Approximation and Error

Theorem (Weierstrass approximation Theorem). Given a function $f \in C[a,b]$ and given $\epsilon > 0$, there exists a polynomial p(x) such that,

$$||f(x) - p|| < \epsilon$$

Using Langrange's recipe to approximate

Take n+1 interpolation points in the [a,b] and collect the function values at all the points. We have n+1 data points. Using Lagrange polynomials, find the interpolant

Theorem (Error equation). Let $f \in C^k[a,b]$, $x_0, x_1, \ldots, x_n \in [a,b]$ and $p \in \mathbb{P}_n$ be the interpolant using these points, then for all x, there exists a $\zeta = \zeta(x) \in (a,b)$ such that

$$f(x) - p(x) = \frac{1}{(n+1)!} f^{(n+1)}(\zeta) \prod_{k=0}^{n} (x - x_k)$$

Note: Here ζ is dependent on the x, i.e, for every x you choose, ζ generally changes.

Proof. Consider the function,

$$\psi(t) = (f(t) - p(t)) \prod_{k=0}^{n} (x - x_k) - (f(x) - p(x)) \prod_{k=0}^{n} (t - x_k)$$

This n+2 roots (n+1 data points and x), applying rolle theorem's gives us that $\phi^{(1)}(t)$ has at least n+1 roots. Applying like this repeatedly on its derivatives, we get that $f^{(n+1)}$ has at least 1 root in [a, b]. Assuming the root to be ζ . We have,

$$f(x) - p(x) = \frac{1}{(n+1)!} f^{(n+1)}(\zeta) \prod_{k=0}^{n} (x - x_k)$$

Approximating the error:

Taking norm on both the sides of error equation, we have,

$$\max_{x \in [a,b]} |f(x) - p(x)| = \frac{1}{(n+1)!} ||f^{(n+1)}(\zeta) \prod_{k=0}^{n} (x - x_k)||$$
 (1)

$$\max_{x \in [a,b]} |f(x) - p(x)| \le \frac{1}{(n+1)!} ||f^{(n+1)}|| \max_{x \in [a,b]} \prod_{k=0}^{n} (x - x_k)$$
 (2)

Chebyshef interpolation points:

$$x_k = \frac{a+b}{2} + \frac{b-a}{2}cos\left(\frac{j\pi}{n}\right)$$

These points minimise $\max_{x \in [a,b]} \prod_{k=0}^{n} (x-x_k)$ and therefore prefered over equally spaced points on real line. These points can be visualised as projections of equally spaced points on the arc of the semicircle with $\frac{a+b}{2}$ as center and $\frac{(b-a)}{2}$ as radius.

1.6 Another method for calculating interpolant

This is similar to the linear algebra method (given as proof(1) to Joseph-Louis Lagrange Theorem) for finding the interpolant.

Consider the polynomial

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

Find the coefficients a_0, a_1, \ldots, a_n by substituting the data points. On substituting x_0 , we get a_0 , again on substituting x_1 and using a_0 , we get a_1 and so on.

1.7 Divided difference - recursion relation

• Divided difference: It is the coefficient of x_n in the interpolant $p \in \mathbb{P}_n$ and denoted by $f[x_0, x_1, \dots, x_n]$.

Using Langrange polynomials, we have

$$p(x) = \sum_{k=0}^{n} f(x_k) \prod_{j=0, j \neq k}^{n} \frac{x - x_j}{x_k - x_j}$$

So the divided difference is

$$f[x_0, x_1, \dots, x_n] = \sum_{k=0}^n f(x_k) \prod_{j=0, j \neq k}^n \frac{1}{x_k - x_j}$$

• interpolant p(x) of x_0, x_1, \ldots, x_n can be calculated using interpolants a(x) and b(x) of x_1, x_2, \ldots, x_n and $x_0, x_1, \ldots, x_{n-1}$ respectively as

$$p(x) = \frac{(x - x_0)a(x) - (x - x_n)b(x)}{x_n - x_0}$$

Theorem (Divided difference recursion theorem).

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

Proof. Let p(x) be the interpolant for x_0, x_1, \ldots, x_m and q(x) be the interpolant for $x_1, x_2, \ldots, x_{m+1}$. Then,

$$L(x) = \frac{(x - x_0)q(x) + (x_{m+1} - x)p(x)}{x_{m+1} - x_0}$$

is an interpolant. Since, interpolant is unique, considering coeff of x_m we have,

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

Theorem (Interpolant using divided differences). Suppose x_0, x_1, \ldots, x_n be the data points. Then interpolant $p \in \mathbb{P}_n$ is

$$p(x) = f[x_0] + f[x_0, x_1](x - x_0) + \dots + f[x_0, x_1, \dots, x_n] \prod_{j=0}^{n-1} (x - x_j)$$

Proof. We prove this by induction. Base case n=0 is trivially satisfied. Assume that this is satisfied for p_k ,

$$p_k(x) = f[x_0] + f[x_0, x_1](x - x_0) + \dots + f[x_0, x_1, \dots, x_k] \prod_{i=0}^{k-1} (x - x_i)$$

Consider the polynomial $p_{k+1}(x) - p_k(x) \in \mathbb{P}_{k+1}$ which has x_0, x_1, \dots, x_k as roots. Hence,

$$p_{k+1}(x) - p_k(x) = c \prod_{j=0}^{k} (x - x_j)$$

Comparing leading coefficient on both sides, we have $c = f[x_0, x_1, \dots, x_k]$. Hence,

$$p_{k+1}(x) = f[x_0] + f[x_0, x_1](x - x_0) + \dots + f[x_0, x_1, \dots, x_{k+1}] \prod_{i=0}^{k} (x - x_i)$$

By PMI,

$$p(x) = f[x_0] + f[x_0, x_1](x - x_0) + \ldots + f[x_0, x_1, \ldots, x_n] \prod_{j=0}^{n-1} (x - x_j)$$

1.8 Time complexity of the algorithms

(Slide is useless)

- Langrange's method $O(n^2)$: Computing each Langrange polynomial can be done in O(n) (Finding the coefficients given roots can be done in $O(n\log^2(n))$). We need to do to this n times. So, O(n).
- Divided differences $O(n^2)$: Summing operations in each stage $n + (n-1) + (n-2) + \ldots + 1$. Hence, $O(n^2)$.
- Divided difference can be considered better because we can extend from n to n+1 and so on without discarding previous computation.

1.9 Weierstrass theorem consequences

• For Weierstrass theorem, take $\epsilon_n = \frac{1}{n}$. Then weirstrass theorem proves the existence of sequence of polynomials $p^{(1)}, p^{(2)}, \ldots$ such that

$$\lim_{n \to \infty} ||f - p^{(n)}|| = 0$$

• If f in not a polynomial, then

$$\lim_{n \to \infty} \text{degree of } p(n) = 0$$

To prove (crude) this, assume that there exist a non-polynomial function f such that $||f - p^{(n)}|| = 0$, then $f = p^{(n)}$ which contradicts the assumtion that f is not a polynomial.

1.10 Spline Interpolation

- Piece wise polynomial: $\phi \in C[a, b]$ is a piecewise polynomial function, if there exists $a = x_0 < x_1 < \ldots < x_n = b$ such that $\phi \in \mathbb{P}_m$ when $x \in [x_i, x_{i+1}]$ for all $i = 0, 1, \ldots, n$ and some m > 0.
- Piece wise polynomial ϕ need not be polynomial in whole domain.
- Splines interpolation for $f \in C[a, b]$
 - Pick some data points x_0, x_1, \ldots, x_n such that $a = x_0 < x_1 < \ldots < x_n = b$
 - Fix m < n
 - Build ϕ in each subinterval $[x_i, x_{i+1}]$ using the following conditions:

$$\phi(x_i) = f_i \qquad \text{for } i = 0, 1, \dots, n$$

$$\lim_{h \to 0+} \phi(x_i - h) = \lim_{h \to 0+} \phi(x_i + h) \qquad \text{for } i = 1, 2, \dots, n-1$$

$$\lim_{h \to 0+} \frac{\mathrm{d}\phi(x_i - h)}{\mathrm{d}x} = \lim_{h \to 0+} \frac{\mathrm{d}\phi(x_i + h)}{\mathrm{d}x} \qquad \text{for } i = 1, 2, \dots, n-1$$

$$\lim_{h \to 0+} \frac{\mathrm{d}^2\phi(x_i - h)}{\mathrm{d}x^2} = \lim_{h \to 0+} \frac{\mathrm{d}^2\phi(x_i + h)}{\mathrm{d}x^2} \qquad \text{for } i = 1, 2, \dots, n-1$$

$$\vdots$$

$$\vdots$$

$$\lim_{h \to 0+} \frac{\mathrm{d}^{n-1}\phi(x_i - h)}{\mathrm{d}x^{n-1}} = \lim_{h \to 0+} \frac{\mathrm{d}^{n-1}\phi(x_i + h)}{\mathrm{d}x^{n-1}} \qquad \text{for } i = 1, 2, \dots, n-1$$

– We have (n+1)+m(n-1)=n(m+1)-(m-1) conditions. We need m-1 more conditions.

Theorem (Error - Linear Splines). Let $f \in C^2[a,b]$ and $s_L(x)$ be the interpolating **linear** spline at (n+1) knots $a = x_0 < x_1 < \ldots < x_n = b$ and let h be the maximum subinterval length, then

$$||f - s_L|| \le \frac{h^2}{8} ||f''||$$

Proof. Consider the interval $[x_i, x_{i+1}]$, $\phi(x)$ is the interpolating polynomial in this interval, using the error equation for interpolating polynomials,

$$f(x) - s_L(x) = \frac{1}{2}f''(\zeta)(x - x_i)(x - x_{i+1})$$

Taking absolute value on both the sides,

$$|f(x) - s_L(x)| = \frac{1}{2} |f''(\zeta)| |(x - x_i)(x - x_{i+1})|$$

$$\leq \frac{1}{2} ||f''|| \frac{h_i^2}{4} \quad \text{where } h_i = \frac{x_{i+1} - x_i}{2}$$

$$\leq \frac{h_i^2}{8} ||f''||$$

Considering $h = max(h_i)$, then for $x \in [a, b]$

$$max_{x \in [a,b]}|f(x) - s_L(x)| \le \frac{1}{8}h^2||f''||$$