

Set 1: Polynomial Interpolation – Part 1

Kyle A. Gallivan

Department of Mathematics

Florida State University

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Interpolation Topics

1. Interpolation Overview
2. Lagrange Interpolation – Section 8.1
3. Newton Interpolation – Section 8.2
4. Complexity and Barycentric Forms – Sections 8.1, 8.2 and 8.3
5. Conditioning and Error Bounds – Sections 8.1 and 8.2
6. Hermite Interpolation – Section 8.5
7. Piecewise Interpolation and Splines – Section 8.4 and 8.8
8. Multidimensional Interpolation – Section 8.6
9. Rational Interpolation (notes)

References

In addition to the text, the following are useful references for this topic.

1. Isaacson and Keller, Analysis of Numerical Methods, Wiley Press, 1966.
2. Bartle, The Elements of Real Analysis, Wiley, Second Edition, 1976.
3. Higham, Accuracy and Stability of Numerical Algorithms, SIAM, Second Edition, 2002.
4. Dahlquist and Bjorck, Numerical Methods, Prentice-Hall, 1974.
5. Ueberhuber, Numerical Computation, Volume 1, Springer, 1995.

Polynomial Interpolation

- Find $p_n(x) \in \mathbb{P}_n | y_i = p_n(x_i) \quad 0 \leq i \leq n$.
- $n + 1$ parameters and $n + 1$ constraints
- global, local nonsmooth, or local smooth polynomial interpolation
- polynomials and their derivatives are cheap to evaluate
- many representations of polynomials, i.e., parameterizations
- efficient interpolation algorithms exist and can be adapted to many circumstances: quadrature, differentiation, integration.
- accuracy of approximation achieved and achievable may be a problem
- accuracy must ultimately be considered in terms of the application exploiting the interpolating polynomial.

Monomial Form, Existence and Uniqueness

Assume the polynomial, $p_n(x)$, is taken in terms of monomials, x^i

$$p_n(x) = \alpha_0 + \alpha_1 x + \alpha_2 x^2 + \cdots + \alpha_n x^n$$

Consider the constraints

$$y_0 = \alpha_0 + \alpha_1 x_0 + \alpha_2 x_0^2 + \cdots + \alpha_n x_0^n$$

$$y_1 = \alpha_0 + \alpha_1 x_1 + \alpha_2 x_1^2 + \cdots + \alpha_n x_1^n$$

$$\vdots$$

$$y_n = \alpha_0 + \alpha_1 x_n + \alpha_2 x_n^2 + \cdots + \alpha_n x_n^n$$

Monomial Form, Existence and Uniqueness

$$\begin{pmatrix} 1 & x_0 & x_0^2 & \dots & x_0^n \\ 1 & x_1 & x_1^2 & \dots & x_1^n \\ \vdots & \vdots & & \ddots & \\ 1 & x_{n-1} & x_{n-1}^2 & \dots & x_{n-1}^n \\ 1 & x_n & x_n^2 & \dots & x_n^n \end{pmatrix} \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \vdots \\ \alpha_{n-1} \\ \alpha_n \end{pmatrix} = \begin{pmatrix} y_0 \\ y_1 \\ \vdots \\ y_{n-1} \\ y_n \end{pmatrix}$$

$$V^T a = y$$

Example

$$(x, y) = \{(1, 10) \ (2, 26) \ (3, 58) \ (4, 112)\}$$

4 distinct x_i implies cubic $p_3(x) = \alpha_0 + \alpha_1 x + \alpha_2 x^2 + \alpha_3 x^3$

$$\alpha_0 + \alpha_1 * 1 + \alpha_2 * 1 + \alpha_3 * 1 = 10$$

$$\alpha_0 + \alpha_1 * 2 + \alpha_2 * 4 + \alpha_3 * 8 = 26$$

$$\alpha_0 + \alpha_1 * 3 + \alpha_2 * 9 + \alpha_3 * 27 = 58$$

$$\alpha_0 + \alpha_1 * 4 + \alpha_2 * 16 + \alpha_3 * 64 = 112$$

Example

$$\begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 4 & 8 \\ 1 & 3 & 9 & 27 \\ 1 & 4 & 16 & 64 \end{pmatrix} \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} 10 \\ 26 \\ 58 \\ 112 \end{pmatrix} \rightarrow \begin{pmatrix} \alpha_0 \\ \alpha_1 \\ \alpha_2 \\ \alpha_3 \end{pmatrix} = \begin{pmatrix} 4 \\ 3 \\ 2 \\ 1 \end{pmatrix}$$

$$\det(V^T) = 12 \text{ and } \kappa_2(V^T) = 1.17 \times 10^3$$

Solution yields: the cubic $p_3(x) = 4 + 3x + 2x^2 + 1x^3$

Complexity: LU factorization $O(n^3)$, Bjorck and Pereyra $O(n^2)$

Monomial Form, Existence and Uniqueness

- The matrix V is called a Vandermonde matrix.
- The determinant is easily computed

$$\det(V) = \det(V^T) = \prod_{j=0}^n \left\{ \prod_{i=j+1}^n (x_i - x_j) \right\}$$

- \therefore if the x_i are $n + 1$ distinct values there is a unique interpolating polynomial of degree n

Vandermonde Systems

- Vandermonde matrices appear in many numerical problems.
- V derived from $x_j = \omega^j$ for $0 \leq j \leq n - 1$ where $\omega = \exp(-2\pi i)/n$, i.e., an n -th root of unity, V/\sqrt{n} is a unitary matrix that defines the Discrete Fourier Transform. It is perfectly conditioned.
- Bjorck and Pereyra have derived and analyzed algorithms to solve $V^T b = c$ and $Vb = c$ requiring $(5n^2)/2$ operations.
- These often produce accurate solutions despite ill-conditioning of V or V^T .
- For interpolation, the algorithm produces the Newton form polynomial as an intermediate result and then converts to the monomial form!

Vandermonde Matrix Conditioning

- Higham (2002) presents a detailed analysis of the conditioning of Vandermonde matrices and the stability of solving related systems.
- V and V^T can be very ill-conditioned.
- The ∞ -norm conditioning for $x_i = 1/(i + 1)$ increases faster than $n!$, i.e., $\kappa_\infty > n^{n+1}$
- for any real choice of x_i , κ_2 increases at least exponentially and

$$\kappa_2 \geq \left(\frac{2}{n+1} \right)^{1/2} \left(1 + \sqrt{2} \right)^{n-1}$$

with equality using equally spaced points on $[0, 1]$.

Vandermonde Matrix Conditioning

- complex roots of unity, i.e., the Fourier matrix V/\sqrt{n} , is the only choice that is perfectly conditioned.
- If $0 \leq x_0 \leq x_1 \leq \dots \leq x_n$ then the computed solution, \hat{a} to $V^T a = y$ using a generalization of Bjorck and Pereyra's algorithm satisfies the (amazing) component-wise bound

$$|y - V^T \hat{a}| \leq (n(n+4)u + O(u^2)) \|V^T\| |\hat{a}|$$

where u is unit roundoff. (Higham 2002)

- So the ill-conditioning does not necessarily preclude an accurate solution when the correct algorithm is chosen and there are some constraints on the x_i .

Alternative Forms

- $p_n(x)$ is unique for distinct x_i
- uniqueness can be proven without using Vandermonde system (consider two polynomials of degree n that agree at $n + 1$ distinct points)
- V^T tends to get increasingly ill-conditioned, $\kappa \sim 10^n$
- other representations can be chosen to construct $p_n(x)$
- Two other forms of interest here:
 1. Lagrange form
 2. Newton form
- As noted, using Bjorck and Pereyra yields the Newton form as an intermediate step so the Vandermonde approach and the monomial form are of limited practical use.

Lagrange Form

Given (x_i, y_i) $0 \leq i \leq n$ with distinct x_i values.

Consider a polynomial $m_0(x) \in \mathbb{P}_n$ defined by

$$m_0(x) = (x - x_1)(x - x_2) \cdots (x - x_{n-1})(x - x_n)$$

$$m_0(x_0) = (x_0 - x_1)(x_0 - x_2) \cdots (x_0 - x_{n-1})(x_0 - x_n)$$

$$\ell_0(x) = \frac{m_0(x)}{m_0(x_0)} = \begin{cases} 1 & \text{if } x = x_0 \\ 0 & \text{if } x = x_j \end{cases}$$

Lagrange Form

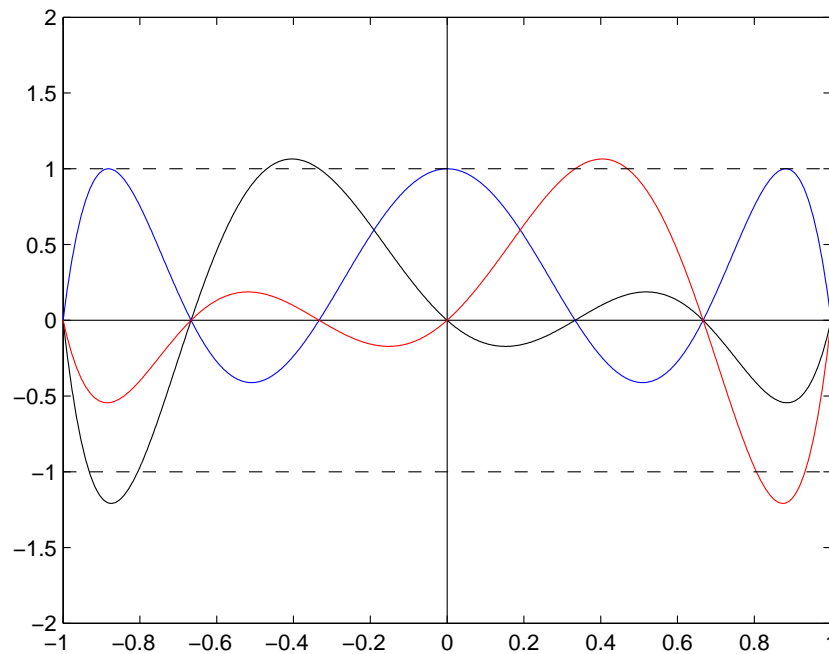
Define

$$m_i(x) = \prod_{j=0, j \neq i}^n (x - x_j) \text{ and } \ell_i(x) = \frac{m_i(x)}{m_i(x_i)} = \begin{cases} 1 & \text{if } x = x_i \\ 0 & \text{if } x = x_j \end{cases}$$

$$p_n(x) = \sum_{i=0}^n y_i \ell_i(x) \rightarrow p_n(x_i) = y_i \quad 0 \leq i \leq n$$

Since the interpolating polynomial is unique this is the same $p_n(x)$ as before.

Characteristic Polynomials



$n = 6$ on $[-1, 1]$, equidistant, $\ell_2(x)$ – black, $\ell_3(x)$ – blue, $\ell_4(x)$ – red

Characteristic Polynomials

- basis for \mathbb{P}_n , i.e., given distinct x_i ,
 $\forall p_n(x) \in \mathbb{P}_n, \quad p_n(x) = \sum_{i=0}^n \alpha_i \ell_i(x)$ with unique $\alpha_0, \dots, \alpha_n$.
- note asymmetric forms
- magnitude is not bounded by 1
- $\ell'_i(x_i)$ is not necessarily 0 (compare location of peak of $\ell_i(x)$ to x_i where $\ell_i(x_i) = 1$)
- matrix form of the interpolation problem is trivial, i.e., $\alpha_i = y_i$ is easily seen.

Example

Let $n = 3$, $(x, y) = \{(1, 10) (2, 26) (3, 58) (4, 112)\}$

$$p_3(x) = 4 + 3x + 2x^2 + 1x^3$$

$$\begin{aligned} p_3(x) &= 10 \times \frac{(x-2)(x-3)(x-4)}{(1-2)(1-3)(1-4)} + 26 \times \frac{(x-1)(x-3)(x-4)}{(2-1)(2-3)(2-4)} \\ &\quad + 58 \times \frac{(x-1)(x-2)(x-4)}{(3-1)(3-2)(3-4)} + 112 \times \frac{(x-1)(x-2)(x-3)}{(4-1)(4-2)(4-3)} \\ &= -\frac{5}{3}(x-2)(x-3)(x-4) + 13(x-1)(x-3)(x-4) \\ &\quad - 29(x-1)(x-2)(x-4) + \frac{56}{3}(x-1)(x-2)(x-3) \end{aligned}$$

Newton Form of Polynomial

Constant: $p_0(x) = y_0$

Point-slope form of line:

$$y(x) = y_0 + m(x - x_0)$$

$$y(x_1) = y_1 = y_0 + m(x_1 - x_0)$$

$$m = \frac{y_1 - y_0}{x_1 - x_0}$$

\Downarrow

$$p_1(x) = y_0 + \frac{y_1 - y_0}{x_1 - x_0}(x - x_0)$$

$$p_1(x_0) = y_0 \text{ and } p_1(x_1) = y_1$$

Newton Form of Polynomial

Define the first divided difference:

$$y[x_0, x_1] = \frac{y_1 - y_0}{x_1 - x_0}$$

The Newton polynomial of degree 1 is then defined as:

$$p_1(x) = y_0 + y[x_0, x_1](x - x_0)$$

Note the form

$$p_1(x) = p_0(x) + q_1(x)$$

Consider the general case of increasing degree of interpolating $p_{n-1}(x)$:

$$p_n(x) = p_{n-1}(x) + q_n(x)$$

Incrementing Newton Form of Polynomial

$$p_n(x) = p_{n-1}(x) + q_n(x) \rightarrow q_n(x) = p_n(x) - p_{n-1}(x)$$

$$q_n(x_i) = p_n(x_i) - p_{n-1}(x_i) = 0 \quad \text{for } 0 \leq i \leq n-1$$

$$\therefore q_n(x) = \alpha_n(x - x_0)(x - x_1) \cdots (x - x_{n-1}) = \alpha_n \omega_n(x)$$

$$p_n(x_n) = y_n = p_{n-1}(x_n) + \alpha_n \omega_n(x_n)$$

\Downarrow

$$\alpha_n = \frac{y_n - p_{n-1}(x_n)}{\omega_n(x_n)} \equiv y[x_0, \dots, x_n]$$

Example

Let $n = 3$, $(x, y) = \{(1, 10) (2, 26) (3, 58) (4, 112)\}$

Recall the monomial form $p_3(x) = 4 + 3x + 2x^2 + 1x^3$

$$p_0(x) = y_0 = 10$$

$$p_1(x) = y_0 + \frac{26 - p_0(2)}{(2 - 1)}(x - 1) = 10 + 16(x - 1)$$

$$p_2(x) = p_1(x) + \frac{58 - p_1(3)}{(3 - 2)(3 - 1)}(x - 1)(x - 2)$$

$$= 10 + 16(x - 1) + 8(x - 1)(x - 2)$$

$$p_3(x) = p_2(x) + \frac{112 - p_2(4)}{(4 - 3)(4 - 2)(4 - 1)}(x - 1)(x - 2)(x - 3)$$

$$= 10 + 16(x - 1) + 8(x - 1)(x - 2) + 1(x - 1)(x - 2)(x - 3)$$

Recursive Form

Given (x_i, y_i) , for $0 \leq i \leq n$, let

- $\omega_{k:k+s}(x) = (x - x_k)(x - x_{k+1}) \cdots (x - x_{k+s})$
- $p_{n-1}(x)$ interpolates y_0, \dots, y_{n-1} and $q_{n-1}(x)$ interpolates y_1, \dots, y_n
- $p_n(x)$ interpolates y_0, \dots, y_n

Lemma.

$$p_n(x) = q_{n-1}(x) + \frac{x - x_n}{x_n - x_0} \left[q_{n-1}(x) - p_{n-1}(x) \right]$$

Recursive Form

Proof. Check interpolation conditions for $1 \leq i \leq n - 1$ and the endpoints.

$$\begin{aligned} p_n(x_i) &= q_{n-1}(x_i) + \frac{x_i - x_n}{x_n - x_0} \left[q_{n-1}(x_i) - p_{n-1}(x_i) \right] \\ &= q_{n-1}(x_i) + \frac{x_i - x_n}{x_n - x_0} \times 0 = y_i \end{aligned}$$

$$\begin{aligned} p_n(x_n) &= q_{n-1}(x_n) + \frac{x_n - x_n}{x_n - x_0} \left[q_{n-1}(x_n) - p_{n-1}(x_n) \right] \\ &= q_{n-1}(x_n) + 0 = y_n \end{aligned}$$

$$\begin{aligned} p_n(x_0) &= q_{n-1}(x_0) + \frac{x_0 - x_n}{x_n - x_0} \left[q_{n-1}(x_0) - p_{n-1}(x_0) \right] \\ &= q_{n-1}(x_0) - q_{n-1}(x_0) + p_{n-1}(x_0) = y_0 \end{aligned}$$

□

Recursive Form

$$p_{n-1}(x) = \sum_{i=0}^{n-1} \omega_{0:i-1}(x) y[x_0, \dots, x_i]$$

$$= y[x_0, \dots, x_{n-1}] x^{n-1} + r_{n-2}(x)$$

$$q_{n-1}(x) = \sum_{i=1}^n \omega_{1:i-1}(x) y[x_1, \dots, x_i]$$

$$= y[x_1, \dots, x_n] x^{n-1} + \tilde{r}_{n-2}(x)$$

$$p_n(x) = \sum_{i=0}^n \omega_{0:i-1}(x) y[x_0, \dots, x_i]$$

$$= y[x_0, \dots, x_n] x^n + \hat{r}_{n-1}(x)$$

Recursive Form

Compare leading term coefficients:

$$p_{n-1}(x) = y[x_0, \dots, x_{n-1}]x^{n-1} + r_{n-2}(x)$$

$$q_{n-1}(x) = y[x_1, \dots, x_n]x^{n-1} + \tilde{r}_{n-2}(x)$$

$$p_n(x) = y[x_0, \dots, x_n]x^n + \hat{r}_{n-1}(x)$$

$$p_n(x) = q_{n-1}(x) + \frac{x - x_n}{x_n - x_0} [q_{n-1}(x) - p_{n-1}(x)]$$

$$= \frac{x^n}{x_n - x_0} [y[x_1, \dots, x_n] - y[x_0, \dots, x_{n-1}]] + \text{lower order terms}$$

$$\therefore y[x_0, \dots, x_n] = \frac{y[x_1, \dots, x_n] - y[x_0, \dots, x_{n-1}]}{x_n - x_0}$$

Example

$$D^k f_i \quad 1 \leq k \leq n, \quad 0 \leq i \leq n - k$$

i	0	1	2	3
x_i	1	2	3	4
f_i	10	26	58	112
$D^1 f_{i-1}$	–	$D^1 f_0 = 16$	$D^1 f_1 = 32$	$D^1 f_2 = 54$
$D^2 f_{i-2}$	–	–	$D^2 f_0 = 8$	$D^2 f_1 = 11$
$D^3 f_{i-3}$	–	–	–	$D^3 f_0 = 1$

$$\text{where } D^k f_i = f[x_i, \dots, x_{i+k}] = \frac{D^{k-1} f_{i+1} - D^{k-1} f_i}{(x_{i+k} - x_i)}$$

$$p_3(x) = f_0 + D^1 f_0 \omega_1(x) + D^2 f_0 \omega_2(x) + D^3 f_0 \omega_3(x)$$

$$p_3(x) = 10 + 16(x - 1) + 8(x - 1)(x - 2) + 1(x - 1)(x - 2)(x - 3)$$

Equidistant Differences

Suppose $x_i = x_{i-1} + h$ for $h > 0$ and $i = 0, \dots$

$$\Delta f(x) = f(x + h) - f(x)$$

$$\nabla f(x) = f(x) - f(x - h)$$

$$\Delta^2 f(x) = \Delta f(x + h) - \Delta f(x)$$

$$\Delta^{m+1} f(x) = \Delta^m f(x + h) - \Delta^m f(x)$$

$$f[x_0, x_1] = \frac{\Delta f(x_0)}{h}$$

$$f[x_0, \dots, x_k] = \frac{\Delta^k f(x_0)}{k!h^k}$$

Equidistant Basic Formulae

$$x = x_0 + sh \rightarrow x - x_i = (s - i)h$$

$$\prod_{i=0}^n (x - x_i) = h^{n+1} \prod_{i=0}^n (s - i) = \pi_n(s) h^{n+1}$$

$\pi_k(s)$ is called the factorial polynomial. We can therefore generalize the binomial coefficient:

$$\binom{s}{k} \equiv \frac{\prod_{i=0}^{k-1} (s - i)}{k!} = \frac{\pi_{k-1}(s)}{k!}$$

Equidistant Newton Polynomial

$$\begin{aligned} & f_0 + f[x_0, x_1](x - x_0) + \dots + f[x_0, \dots, x_n](x - x_0) \cdots (x - x_{n-1}) \\ &= f_0 + \frac{\Delta f_0}{h}(x - x_0) + \dots + \frac{\Delta^n f_0}{n!h^n}(x - x_0) \cdots (x - x_{n-1}) \\ &= f_0 + \frac{\Delta f_0}{h}\pi_0(s)h + \dots + \frac{\Delta^n f_0}{n!h^n}\pi_{n-1}(s)h^n \\ &= f_0 + \frac{\Delta f_0}{1!}\pi_0(s) + \dots + \frac{\Delta^n f_0}{n!}\pi_{n-1}(s) \\ &= f_0 + \binom{s}{1}\Delta f_0 + \dots + \binom{s}{i}\Delta^i f_0 + \dots + \binom{s}{n}\Delta^n f_0 \end{aligned}$$