

# Spline Interpolation (I) and Hermite wrap-up

Wednesday, October 16, 2024 11:54 AM

Class 22: October 16, 2024

**Recall:** Last time, we proposed an extension to the polynomial interpolation problem on  $n+1$  points (nodes)  $x_0, x_1, \dots, x_n$ . In this extension, we provide function data  $y_i = f(x_i)$  and derivative data  $z_i = f'(x_i)$  at these nodes, resulting in  $2n+2$  data points. The Hermite interpolant is the unique polynomial of degree  $\leq 2n+1$  that satisfies all of these conditions (so,  $h(x_i) = y_i$  and  $h'(x_i) = z_i$  for  $i=0, 1, \dots, n$ ).

We briefly touched on the following:

1. We can build a Vandermonde matrix for this problem, known as the Confluent-Vandermonde. Its determinant is non-zero if and only if the  $x_i$ 's are distinct. Its condition number is terrible (grows exponentially with  $n$ ).
2. We can build a Hermite-Lagrange basis  $\{H_j(x), K_j(x)\}$ , consisting of polynomials of degree  $\leq 2n+1$  that satisfy one condition = 1, and the rest = 0. The interpolant is then the sum of  $y_j * H_j(x) + z_j * K_j(x)$ , and the formulas for both depend on  $M_j(x) = L_j(x)^2$
3. We can build a Hermite-Newton basis, and come up with a process to compute a Hermite-Newton tableau and then use Horner to evaluate  $h(x)$ . This requires us to "repeat" each  $x_i$  twice, and define  $f[x_i, x_i] = f'(x_i)$ .
4. Finally, we mentioned that the same tricks for Barycentric Lagrange can be played to find a 2nd barycentric formula for Hermite-Lagrange.

Today, we will talk about error estimates for Hermite, wrapping up this subject. We will then move on to talk about an extremely useful interpolation method: the use of "splines" (a kind of piecewise polynomial). Splines are ubiquitous in curve fitting, approximation, computer graphics (e.g. Pixar) and scientific computing.

## ERROR EST FOR HERMITE:

FOR INT ON  $n+1$  points given  $(x_i, y_i)$  data.

Assumed  $y_i = f(x_i)$  (Exact),  $f \in C^{n+1}([a, b])$

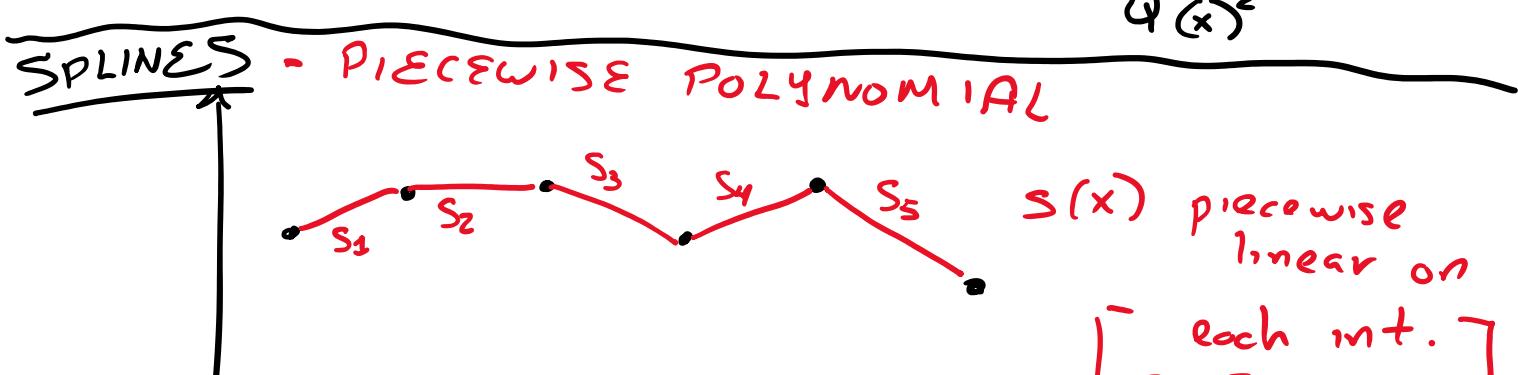
$[a, b]$  contains  $x_0, x_1, \dots, x_n$ .

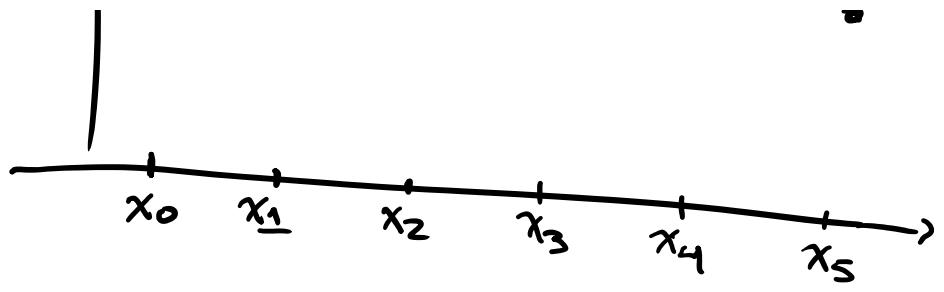
$$f(x) - p(x) = \frac{f^{(n+1)}(x)}{(n+1)!} \underbrace{(x-x_0)(x-x_1) \dots (x-x_n)}_{\psi(x)}$$

## HERMITE:

Assume:  $y_i = f(x_i)$ ,  $z_i = f'(x_i)$ ,  $f \in C^{2n+2}([a, b])$

$$f(x) - h(x) = \frac{f^{(2n+2)}(x)}{(2n+2)!} \underbrace{(x-x_0)^2(x-x_1)^2 \dots (x-x_n)^2}_{\psi(x)^2}$$





$$x_0 < x_1 < x_2 < \dots < x_5 < \dots$$

each int.

$$\boxed{I_i = [x_{i-1}, x_i] \quad i=1, \dots, n}$$

$$\boxed{S_i(x) = s(x) \quad |_{I_i}}$$

$$S_i \in \mathcal{D}_1$$

①  $s(x)$  is continuous.

$$S_1(x_1) = S_2(x_1)$$

$$S_2(x_2) = S_3(x_2)$$

$$\vdots$$

$$\{S_i(x_i) = S_{i+1}(x_i)\}_{i=1}^{n-1}$$

FORMULA FOR  $S_i(x)$ :

Given data  $(x_i, y_i)$   $i=0, \dots, n$

$S_i$  is defined by  $(x_{i-1}, y_{i-1})$  and  $(x_i, y_i)$

$$S_i(x) = y_{i-1} \left( \frac{x - x_i}{x_{i-1} - x_i} \right) + y_i \left( \frac{x - x_{i-1}}{x_i - x_{i-1}} \right)$$

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

$$\boxed{S_i(x) = \frac{1}{h_i} (y_{i-1}(x_i - x) + y_i(x - x_{i-1}))}$$

Given  $(x_i, y_i)$  and  $x_{\text{eval}}$  (m pts)

$$\rightarrow S(x_{\text{eval}})$$

ERROR ANALYSIS?

QUESTION.

$$f(x) - s_i(x) = ?$$

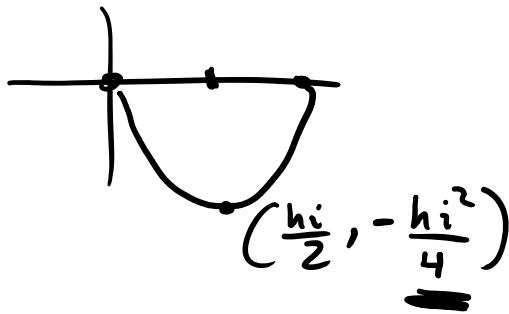
$$|f(x) - s_i(x)| \leq ?$$

Assume  $f \in C^2([a, b])$ ,  $x \in I_i = [x_{i-1}, x_i]$  with width  $h_i$ .

$$f(x) - s_i(x) = \frac{f''(n_x)}{2!} (x - x_{i-1})(x - x_i)$$

$$|f(x) - s_i(x)| \leq \underbrace{\max_{x \in [a, b]} |f''(x)|}_{2} \underbrace{\max_{x \in I_i} |(x - x_{i-1})(x - x_i)|}_{\|f''\|_\infty}$$

$$\max_{x \in [0, h]} |x(x-h)|$$



$$\begin{aligned}s_i(x) &= x(x - h_i) \\ &= x^2 - h_i x\end{aligned}$$

$$s'_i(x) = 2x - h_i \rightarrow x = \frac{h_i}{2}$$

$$|f(x) - s_i(x)| \leq \frac{\|f''\|_\infty}{2} \cdot \left(\frac{h_i^2}{4}\right) \rightarrow O(h^2)$$

Partition is equispaced  $h_i = h$ .  $n \rightarrow \infty$ ,  $h \rightarrow 0$

## CUBIC SPLINES

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$$S_i(x) = S(x)|_{I_i} \in \mathcal{D}_3.$$

- ① Continuity.
- ②  $s', s'' \rightarrow$  Continuity.
- ③ Space of PW CUBIC  $\rightarrow$  problem is well-defined.