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## Numerical Analysis Hilary Term 2021

### Lecture 1: Lagrange Interpolation

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Numerical analysis is the study of computational algorithms for solving problems in scientific computing. It combines mathematical beauty, rigor and numerous applications; we hope you'll enjoy it! In this course we will cover the basics of three key fields in the subject:

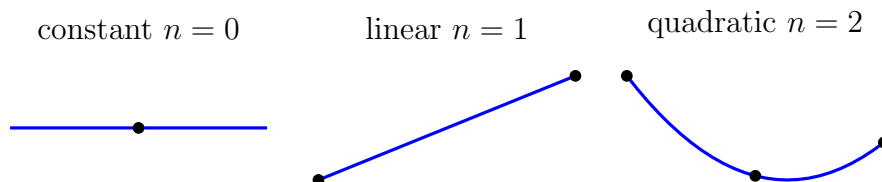
- Approximation Theory (lectures 1, 9–11); recommended reading: L. N. Trefethen, *Approximation Theory and Approximation Practice*, and E. Süli and D. F. Mayers, *An Introduction to Numerical Analysis*.
- Numerical Linear Algebra (lectures 2–8); recommended reading: L. N. Trefethen and D. Bau, *Numerical Linear Algebra*.
- Numerical Solution of Differential Equations (lectures 12–16); recommended reading: E. Süli and D. F. Mayers, *An Introduction to Numerical Analysis*.

This first lecture comes from Chapter 6 of Süli and Mayers.

**Notation:**  $\Pi_n = \{\text{real polynomials of degree } \leq n\}$

**Setup:** Given data  $f_i$  at distinct  $x_i$ ,  $i = 0, 1, \dots, n$ , with  $x_0 < x_1 < \dots < x_n$ , can we find a polynomial  $p_n$  such that  $p_n(x_i) = f_i$ ? Such a polynomial is said to **interpolate** the data, and (as we shall see) can approximate  $f$  at other values of  $x$  if  $f$  is smooth enough. This is the most basic question in approximation theory.

**E.g.:**



**Theorem.**  $\exists p_n \in \Pi_n$  such that  $p_n(x_i) = f_i$  for  $i = 0, 1, \dots, n$ .

**Proof.** Consider, for  $k = 0, 1, \dots, n$ , the “cardinal polynomial”

$$L_{n,k}(x) = \frac{(x - x_0) \cdots (x - x_{k-1})(x - x_{k+1}) \cdots (x - x_n)}{(x_k - x_0) \cdots (x_k - x_{k-1})(x_k - x_{k+1}) \cdots (x_k - x_n)} \in \Pi_n. \quad (1)$$

Then  $L_{n,k}(x_i) = \delta_{ik}$ , that is,

$$L_{n,k}(x_i) = 0 \text{ for } i = 0, \dots, k-1, k+1, \dots, n \text{ and } L_{n,k}(x_k) = 1.$$

So now define

$$p_n(x) = \sum_{k=0}^n f_k L_{n,k}(x) \in \Pi_n \quad (2)$$

$\implies$

$$p_n(x_i) = \sum_{k=0}^n f_k L_{n,k}(x_i) = f_i \text{ for } i = 0, 1, \dots, n. \quad \square$$

The polynomial (2) is the **Lagrange interpolating polynomial**.

**Theorem.** The interpolating polynomial of degree  $\leq n$  is unique.

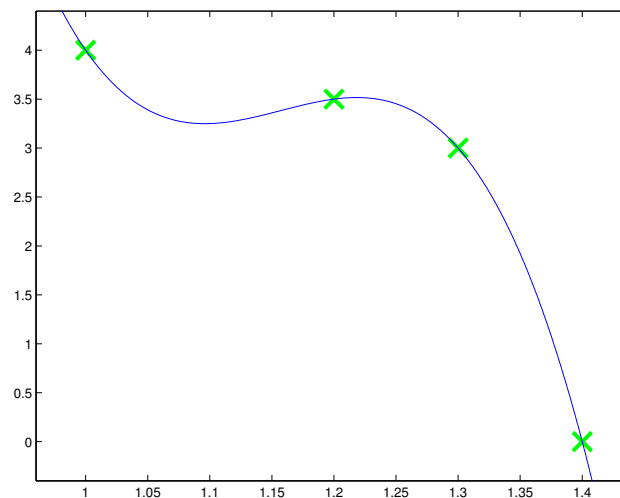
**Proof.** Consider two interpolating polynomials  $p_n, q_n \in \Pi_n$ . Their difference  $d_n = p_n - q_n \in \Pi_n$  satisfies  $d_n(x_k) = 0$  for  $k = 0, 1, \dots, n$ . i.e.,  $d_n$  is a polynomial of degree at most  $n$  but has at least  $n + 1$  distinct roots. Algebra  $\implies d_n \equiv 0 \implies p_n = q_n$ .  $\square$

**Matlab:**

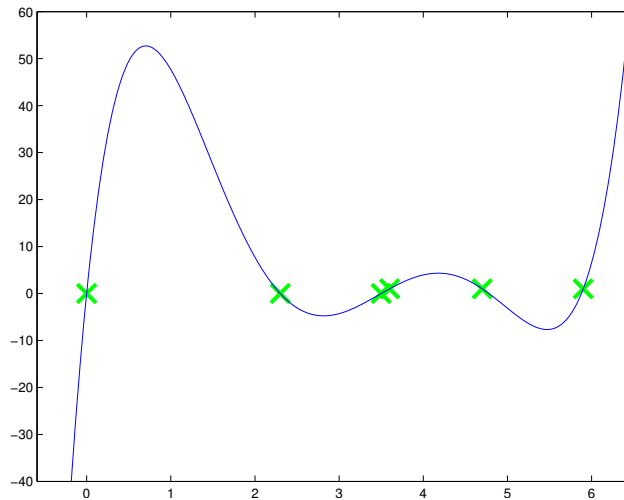
```
>> help lagrange
```

```
LAGRANGE Plots the Lagrange polynomial interpolant for the  
given DATA at the given KNOTS
```

```
>> lagrange([1,1.2,1.3,1.4],[4,3.5,3,0]);
```



```
>> lagrange([0,2.3,3.5,3.6,4.7,5.9],[0,0,0,1,1,1]);
```



**Data from an underlying smooth function:** Suppose that  $f(x)$  has at least  $n + 1$  smooth derivatives in the interval  $(x_0, x_n)$ . Let  $f_k = f(x_k)$  for  $k = 0, 1, \dots, n$ , and let  $p_n$  be the Lagrange interpolating polynomial for the data  $(x_k, f_k)$ ,  $k = 0, 1, \dots, n$ .

**Error:** How large can the error  $f(x) - p_n(x)$  be on the interval  $[x_0, x_n]$ ?

**Theorem.** For every  $x \in [x_0, x_n]$  there exists  $\xi = \xi(x) \in (x_0, x_n)$  such that

$$e(x) \stackrel{\text{def}}{=} f(x) - p_n(x) = (x - x_0)(x - x_1) \cdots (x - x_n) \frac{f^{(n+1)}(\xi)}{(n + 1)!}, \quad (3)$$

where  $f^{(n+1)}$  is the  $(n + 1)$ -st derivative of  $f$ .

**Proof.** Trivial for  $x = x_k$ ,  $k = 0, 1, \dots, n$  as  $e(x) = 0$  by construction. So suppose  $x \neq x_k$ . Let

$$\phi(t) \stackrel{\text{def}}{=} e(t) - \frac{e(x)}{\pi(x)} \pi(t),$$

where

$$\begin{aligned} \pi(t) &\stackrel{\text{def}}{=} (t - x_0)(t - x_1) \cdots (t - x_n) \\ &= t^{n+1} - \left( \sum_{i=0}^n x_i \right) t^n + \cdots + (-1)^{n+1} x_0 x_1 \cdots x_n \\ &\in \Pi_{n+1}. \end{aligned}$$

Now note that  $\phi$  vanishes at  $n + 2$  points  $x$  and  $x_k$ ,  $k = 0, 1, \dots, n$ .  $\implies \phi'$  vanishes at  $n + 1$  points  $\xi_0, \dots, \xi_n$  between these points  $\implies \phi''$  vanishes at  $n$  points between these new points, and so on until  $\phi^{(n+1)}$  vanishes at an (unknown) point  $\xi$  in  $(x_0, x_n)$ . But

$$\phi^{(n+1)}(t) = e^{(n+1)}(t) - \frac{e(x)}{\pi(x)} \pi^{(n+1)}(t) = f^{(n+1)}(t) - \frac{e(x)}{\pi(x)} (n + 1)!$$

since  $p_n^{(n+1)}(t) \equiv 0$  and because  $\pi(t)$  is a monic polynomial of degree  $n + 1$ . The result then follows immediately from this identity since  $\phi^{(n+1)}(\xi) = 0$ .

□

**Example:**  $f(x) = \log(1 + x)$  on  $[0, 1]$ . Here,  $|f^{(n+1)}(\xi)| = n!/(1 + \xi)^{n+1} < n!$  on  $(0, 1)$ . So  $|e(x)| < |\pi(x)|n!/(n + 1)! \leq 1/(n + 1)$  since  $|x - x_k| \leq 1$  for each  $x, x_k$ ,  $k = 0, 1, \dots, n$ , in

$[0, 1] \implies |\pi(x)| \leq 1$ . This is probably pessimistic for many  $x$ , e.g. for  $x = \frac{1}{2}$ ,  $\pi(\frac{1}{2}) \leq 2^{-(n+1)}$  as  $|\frac{1}{2} - x_k| \leq \frac{1}{2}$ .

This shows the important fact that the error can be large at the end points when samples  $\{x_k\}$  are equispaced points, an effect known as the ‘‘Runge phenomena’’ (Carl Runge, 1901), which we return to in lecture 4.

**Generalisation:** Given data  $f_i$  and  $g_i$  at distinct  $x_i$ ,  $i = 0, 1, \dots, n$ , with  $x_0 < x_1 < \dots < x_n$ , can we find a polynomial  $p$  such that  $p(x_i) = f_i$  and  $p'(x_i) = g_i$ ? (i.e., interpolate derivatives in addition to values)

**Theorem.** There is a unique polynomial  $p_{2n+1} \in \Pi_{2n+1}$  such that  $p_{2n+1}(x_i) = f_i$  and  $p'_{2n+1}(x_i) = g_i$  for  $i = 0, 1, \dots, n$ .

**Construction:** Given  $L_{n,k}(x)$  in (1), let

$$\begin{aligned} H_{n,k}(x) &= [L_{n,k}(x)]^2(1 - 2(x - x_k)L'_{n,k}(x_k)) \\ \text{and } K_{n,k}(x) &= [L_{n,k}(x)]^2(x - x_k). \end{aligned}$$

Then

$$p_{2n+1}(x) = \sum_{k=0}^n [f_k H_{n,k}(x) + g_k K_{n,k}(x)] \quad (4)$$

interpolates the data as required. The polynomial (4) is called the **Hermite interpolating polynomial**. Note that  $H_{n,k}(x_i) = \delta_{ik}$  and  $H'_{n,k}(x_i) = 0$ , and  $K_{n,k}(x_i) = 0$ ,  $K'_{n,k}(x_i) = \delta_{ik}$ .

**Theorem.** Let  $p_{2n+1}$  be the Hermite interpolating polynomial in the case where  $f_i = f(x_i)$  and  $g_i = f'(x_i)$  and  $f$  has at least  $2n+2$  smooth derivatives. Then, for every  $x \in [x_0, x_n]$ ,

$$f(x) - p_{2n+1}(x) = [(x - x_0)(x - x_1) \cdots (x - x_n)]^2 \frac{f^{(2n+2)}(\xi)}{(2n+2)!},$$

where  $\xi \in (x_0, x_n)$  and  $f^{(2n+2)}$  is the  $(2n+2)$ nd derivative of  $f$ .

Proof (non-examinable): see Sili and Mayers, Theorem 6.4. □

We note that as  $x_k \rightarrow 0$  in (3), we essentially recover Taylor’s theorem with  $p_n(x)$  equal to the first  $n+1$  terms in Taylor’s expansion. Taylor’s theorem can be regarded as a special case of Lagrange interpolation where we interpolate high-order derivatives at a single point.