

it makes sense to look for poly of degree ≤ 3 since this forms a 4 dimensional subspace spanned by the monomials $\{1, x, x^2, x^3\}$. However it simplifies life to consider another basis:

$$P(x) = a + b(x-x_0) + c(x-x_0)^2 + d(x-x_0)^2(x-x_1)$$

$$P'(x) = b + 2c(x-x_0) + 2d(x-x_0)(x-x_1) + d(x-x_0)^2$$

Writing interpolation conditions we get:

$$f(x_0) = a$$

$$f'(x_0) = b$$

$$f(x_1) = a + b(x_1-x_0) + c(x_1-x_0)^2$$

$$f'(x_1) = b + 2c(x_1-x_0) + d(x_1-x_0)^2$$

a, b easy to get. then c follows and then d

→ problem is solvable for any prescribe values of $f(x_i), f'(x_i)$.

However the general problem where some derivatives are missing doesn't need to be solvable and is called Birkhoff interp.

Here is an example of the possible subtleties one can encounter.

Example: Find a poly st. $p(0) = 0, p'(\frac{1}{2}) = 2, p(1) = 1$.

Three conditions \Rightarrow natural choice is:

$$p(x) = a + bx + cx^2, \quad p'(x) = b + 2cx$$

$$p(0) = 0 \Rightarrow a = 0 \quad \text{other conditions become: } \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} b \\ c \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

↑ singular matrix.

Try cubic:

$$p(x) = a + bx + cx^2 + dx^3$$
$$p'(x) = b + 2cx + 3dx^2$$

and equations become:

we get $a=0$

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$$\begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & \frac{3}{4} \end{bmatrix} \begin{bmatrix} b \\ c \\ d \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

which has infinitely many sol:

$$\begin{pmatrix} b \\ c \\ d \end{pmatrix} = \begin{bmatrix} -4 \\ 5 \\ 0 \end{bmatrix} + k \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$$

↑
some const.

Hermite interpolation conditions (which give uniquely solvable problems)

$$(HIC) \quad p^{(j)}(x_i) = c_{ij} \quad \text{for } 0 \leq j \leq k_i - 1$$
$$0 \leq i \leq m$$

how many conditions do we have in total?

$$k_1 + k_2 + k_3 + \dots + k_m = m + 1$$

↓
degree of poly we will use
to interpolate.

note: # of derivatives can vary from node to node.

However if a derivative is known at a node all previous (lower order) derivatives need to be known as well.

Theorem (on Hermite interp) There is a unique poly of degree $\leq m = k_1 + k_2 + \dots + k_m - 1$ satisfying (HIC).

Proof: Assume there are two polynomials p, q of degree $\leq m$ satisfying (HiC). We will show that $p \equiv q$, i.e. that (HiC) uniquely determines a poly of degree $\leq m$.

Let $r(x) = p(x) - q(x)$, by (HiC):

$$r^{(j)}(x_i) = 0 \quad \text{for } i=0, \dots, n \text{ and } j=0, \dots, k_i-1$$

This means x_i is a zero of multiplicity k_i of r ; which gives us a total of $k_1 + k_2 + \dots + k_n = m+1$ zeros.

Since $r(x)$ is poly of degree $\leq m$ with $m+1$ zeros:

$$r \equiv 0 \text{ and } p \equiv q. \quad \text{QED.}$$

Example: what if we did Hermite interp at only one node?

$$p^{(j)}(x_0) = c_{0j}, \quad j=0, 1, \dots, k$$

$$P(x) = \text{Taylor poly of degree } k$$

$$= c_{00} + c_{01}(x-x_0) + \frac{c_{02}}{2!}(x-x_0)^2 + \dots + \frac{c_{0k}}{k!}(x-x_0)^k$$

Hermite interpolation using Newton divided differences

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First a simple example:

(HIC): $p(x_0) = c_{00}$, $p'(x_0) = c_{01}$, $p(x_1) = c_{10}$

(Hermite interp poly will be a quadratic)

x_0	c_{00}	c_{01}	?
x_0	c_{01}	?	
x_1	c_{10}		

why do we say $f[x_0, x_0] = f'(x_0)$?

well recall:

$$f[x_0, x] = \frac{f(x) - f(x_0)}{x - x_0}$$

thus $\lim_{x \rightarrow x_0} f[x_0, x] = f'(x_0)$

→ Hermite conditions give entries in divided difference table.

a node is repeated as many times as there are deriv. specified.

Back to example we started with:

$$\begin{aligned} p(x_0) &= f(x_0) & \text{and} & & p(x_1) &= f(x_1) \\ p'(x_0) &= f'(x_0) & & & p'(x_1) &= f'(x_1) \end{aligned}$$

we can write table:

x_0	$f(x_0)$	$f'(x_0)$	$f[x_0, x_0, x_1]$	$f[x_0, x_0, x_1, x_1]$
x_0	$f(x_0)$	$f[x_0, x_1]$	$f[x_0, x_1, x_1]$	
x_1	$f(x_1)$	$f'(x_1)$		
x_1	$f(x_1)$			

and

$$p(x) = f(x_0) + f'(x_0)(x - x_0) + f[x_0, x_0, x_1](x - x_0)^2$$

(by reading first row)

$$+ f[x_0, x_0, x_1, x_1](x - x_0)^2(x - x_1)$$

How to interpret divided differences with more than 2 repetitions?

Recall last theorem of interp section:

$$f[x_0, x_1, x_2, \dots, x_m] = \frac{f^{(m)}(\xi)}{m!}$$

where the ξ lies in some interval containing x_0, \dots, x_m .

Now shrink interval to a single point (x_0):

$$f[\underbrace{x_0, x_0, \dots, x_0}_{n+1 \text{ times}}] = \frac{f^{(n)}(x_0)}{n!}$$

Example:

x	1	2
$P(x)$	2	6
$P'(x)$	3	7
$P''(x)$		8

1	2	3	?	?	?
1	2	?	?	?	
2	6	7	(4)		→ why?
2	6	7			
2	6				

Calculating missing entries gives:

1	2	(3)	→ 1	→ 2	→ -1
1	2	4	→ 3	→ 1	
2	6	(7)	(4)		
2	6	(7)			
2	6				

Thus the interp poly is:

$$P(x) = 2 + 3(x-1) + 1 \cdot (x-1)^2 + 2(x-1)^2(x-2) - 1 \cdot (x-1)^2(x-2)^2$$

Divided differences with repetitions:

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Def interpolation with repeated nodes

We say f interpolates 0 at x_0, x_1, \dots, x_m if $f^{(k-1)}(x_j) = 0$ for each point that occurs k or more times in the list x_0, x_1, \dots, x_m .

Example: f interpolates 0 at the points 1, 3, 4, 1, 3, 1, 10 if:

$$f(1) = f'(1) = f''(1) = 0$$

$$f(3) = f'(3) = 0$$

$$f(4) = 0$$

$$f(10) = 0$$

Note: p interpolates 0 at x_0, x_1, \dots, x_m iff

$$p(x) = \prod_{j=0}^m (x - x_j)^{\alpha_j} \underbrace{r(x)}_{\text{poly}}$$

Def: f interp g at x_0, \dots, x_m if $f-g$ interp. 0 at x_0, \dots, x_m .

We can rephrase the theorem on uniqueness of interpolation to nodes with repetition as follows:

Theorem: Let x_0, x_1, \dots, x_m be a list of points with no element repeated more than k times, and let f be a $C^{k-1} [a, b]$ function ($[a, b]$ contains points). Then there is a unique polynomial of degree $\leq m$ interpolating f at x_0, x_1, \dots, x_m .

Note: $f[x_0, x_1, \dots, x_m] \equiv$ coeff of x^m in poly of degree $\leq m$ interp f at x_0, x_1, \dots, x_m .

if ξ appears k times in list then $f^{(k-1)}(\xi)$ must exist. if not then divided difference does not exist.

Theorem (General Newton interp poly)

If f is sufficiently diffble so that divided differences below exist then the unique poly of degree $\leq m$ interp. f at x_0, x_1, \dots, x_m (allowing repetitions) is given by:

$$p(x) = \sum_{j=0}^m f[x_0, x_1, \dots, x_j] \prod_{i=0}^{j-1} (x-x_i)$$

(by convention: $\prod_{i=0}^{-1} (x-x_i) = 1$.)

proof: by induction

$n=0$: $f[x_0]$ is constant poly interp f at x_0

Induction hypothesis: assume

$$q(x) = \sum_{j=0}^{n-1} f[x_0, x_1, \dots, x_j] \prod_{i=0}^{j-1} (x-x_i)$$

interpolates f at nodes x_0, x_1, \dots, x_{n-1} (i.e. theorem is true for $n-1$ nodes)

let $p(x)$ be poly interp f at x_0, \dots, x_n . We know such poly exists and is unique. By def of div. diff:

$$f[x_0, x_1, \dots, x_n] = \text{coeff. in front of } x^n \text{ in } p$$

Therefore:

$$r(x) = p(x) - f[x_0, \dots, x_m] \prod_{i=0}^{m-1} (x - x_i)$$

is of degree $\leq m-1$

and it interpolates f at the nodes x_0, x_1, \dots, x_{m-1} . (Why?)

$$\Rightarrow q(x) = p(x) - f[x_0, \dots, x_m] \prod_{i=0}^{m-1} (x - x_i)$$

$$\Rightarrow p(x) = \sum_{i=0}^m f[x_0, \dots, x_i] \prod_{j=0}^{i-1} (x - x_j) \quad \text{QED.}$$

Question: Assume wlog that $x_0 \leq x_1 \leq \dots \leq x_m$. Do we have a formula for:

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

when repetitions are allowed?

obviously there is a problem with $x_m = x_0$ but then:

$$f[\underbrace{x_0, x_0, \dots, x_0}_{m+1 \text{ times}}] = \frac{1}{m!} f^{(m)}(x_0)$$

Thus we get the theorem (stated w/o proof here)

Then (recursive formula for divided differences)

Let $x_0 \leq x_1 \leq \dots \leq x_m$. Then

$$f[x_0, x_1, \dots, x_m] = \begin{cases} \frac{f[x_1, x_2, \dots, x_m] - f[x_0, \dots, x_{m-1}]}{x_m - x_0} & \text{if } x_m \neq x_0 \\ \frac{1}{m!} f^{(m)}(x_0) & \text{if } x_m = x_0 \end{cases}$$

note: ordering of nodes is here only to simplify test "all nodes are the same".

§3.4 Cubic Spline interpolation

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Big problems with polynomial interpolation: can come when we increase the number of points or the degree of the interpolating polynomial:

- the higher the degree of a polynomial, the more oscillatory it is (in general)
- the interpolation error depends on $n+1$ -th derivative of function. So in order for poly interp to work well we need a very smooth underlying function, which is not always the case.

Splines \equiv piecewise polynomial interp: ie. we subdivide interval in many subintervals and interpolate w/ polynomial inside each subinterval, enforcing some continuity conditions.

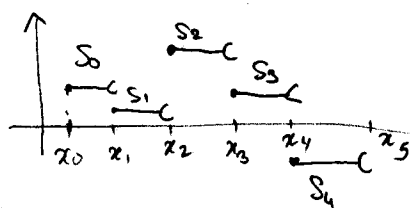
A spline of degree k having $n+1$ distinct knots (nodes) x_0, x_1, \dots, x_n is a function $S(x)$ s.t.

- On $[x_{i-1}, x_i)$, S is a polynomial of degree $\leq k$.
- S has continuous $k-1$ derivatives on $[x_0, x_n]$.

Examples:

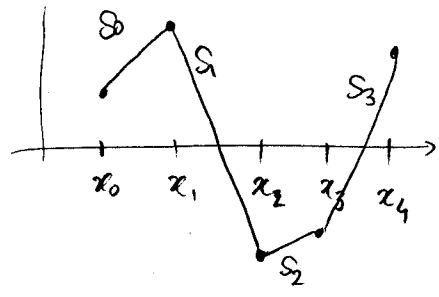
Spline of degree 0: piecewise constant

$$S(x) = \begin{cases} S_0(x) = c_0 & x \in [x_0, x_1) \\ S_1(x) = c_1 & x \in [x_1, x_2) \\ \vdots \\ S_{n-1}(x) = c_{n-1} & x \in [x_{n-1}, x_n) \end{cases}$$



Spline of degree 1:

$$S(x) = \begin{cases} S_0(x) = a_0x + b_0 & [x_0, x_1] \\ S_1(x) = a_1x + b_1 & [x_1, x_2] \\ \vdots & \vdots \\ S_{n-1}(x) = a_{n-1}x + b_{n-1} & [x_{n-1}, x_n] \end{cases} \text{ piecewise linear interp}$$



Spline of degree 2? can construct quadratics on each subinterval s.t. values match, however there are not enough degrees of freedom to enforce continuity of S' at the knots.

cubic splines: are the most commonly used spline and are "optimal" in a sense we shall see later. Assume we are given a table:

x	x_0	x_1	x_2	\dots	x_m	\leftarrow knots
y	y_0	y_1	y_2	\dots	y_m	\leftarrow values at knots.

On each subinterval $[x_i, x_{i+1}]$, $S(x) \equiv$ cubic function $S_i(x)$.

$$S(x) = \begin{cases} S_0(x) & x \in [t_0, t_1] \\ S_1(x) & x \in [t_1, t_2] \\ \vdots & \vdots \\ S_{n-1}(x) & x \in [t_{n-1}, t_n] \end{cases}$$

with the conditions:

- i) $S_j(x_j) = y_j$ and $S_j(x_{j+1}) = y_{j+1}$, $j = 0, \dots, n-1$
- ii) $S'_{j+1}(x_{j+1}) = S'_j(x_{j+1})$, $j = 0, \dots, n-2$
- iii) $S''_{j+1}(x_{j+1}) = S''_j(x_{j+1})$, $j = 0, \dots, n-2$

or in words:

i) = continuity at the knots = $2n$ conditions
(2 per subinterval)

ii) = continuity of derivatives at all internal knots = $n-1$ conditions

iii) = second deriv. = $n-1$ conditions

TOTAL = $4n-2$ equations

unknowns = $4 \times n$ (since a cubic per interval can be def using four parameters).

→ we need 2 more conditions, which are usually specified by:

a) $S''(x_0) = S''(x_n) = 0$ = free or natural boundary

b) $S'(x_0) = f'(x_0), S'(x_n) = f'(x_n)$ = clamped boundary.

"Spline" is actually a term used in shipyards for thin wood strips that were used to draw the shape of the hull.

► Condition a) gives the shape of a strip of wood constrained to passing through certain points and left free at the ends.

► condition b) means the spline is clamped at its end points. (but requires knowledge of derivative of function, i.e. strip direction is known at end points).

Spline construction

$$S_j(x) = a_j + b_j(x-x_j) + c_j(x-x_j)^2 + d_j(x-x_j)^3$$

$j = 0 \dots n-1$

function values (condi)

$S_j(x_j) = a_j = y_j \quad j = 0 \dots n-1$

We also need $S_j(x_{j+1}) = y_{j+1}$, $j = 0 \dots n-1$ i.e.

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$$\boxed{a_j + b_j h_j + c_j h_j^2 + d_j h_j^3 = y_{j+1}} \quad j = 0 \dots n-1 \quad (*)$$

where $h_j = x_{j+1} - x_j$ to simplify notation.

derivatives (cond ii)

$$S_j'(x) = b_j + 2c_j(x - x_j) + 3d_j(x - x_j)^2$$

$$S_j'(x_{j+1}) = S_{j+1}'(x_{j+1}), \quad j = 0, \dots, n-2$$

$$\Leftrightarrow \boxed{b_{j+1} = b_j + 2c_j h_j + 3d_j h_j^2} \quad j = 0 \dots n-2 \quad (**)$$

second derivatives (cond iii)

$$S_j''(x) = 2c_j + 6d_j(x - x_j)$$

$$S_j''(x_{j+1}) = S_{j+1}''(x_{j+1}), \quad j = 0, \dots, n-2$$

$$\Leftrightarrow \boxed{2c_{j+1} = 2c_j + 6d_j h_j} \quad (**')$$

Solving for d_j we get: $\boxed{d_j = \frac{1}{3h_j}(c_{j+1} - c_j)}$ replacing in (*) and (**)

$$\Rightarrow \begin{cases} a_j + b_j h_j + \frac{h_j^2}{3}(2c_j + c_{j+1}) = a_{j+1} & (1) \\ b_j + h_j(c_j + c_{j+1}) = b_{j+1} & (2) \end{cases}$$

Solving for b_j in (1):

$$b_j = \frac{1}{h_j} (a_{j+1} - a_j) - \frac{h_j}{3} (2c_j + c_{j+1}) \quad (3)$$

for $j = 0, \dots, m-1$

replacing in (2):

$$\frac{1}{h_{j+1}} (a_{j+2} - a_{j+1}) - \frac{h_{j+1}}{3} (2c_{j+1} + c_{j+2}) = \frac{1}{h_j} (a_{j+1} - a_j) - \frac{h_j}{3} (2c_j + c_{j+1}) \\ = h_j (c_j + c_{j+1})$$

$$h_j c_j + 2(h_j + h_{j+1})c_{j+1} + h_{j+1} c_{j+2} = \frac{3}{h_{j+1}} (a_{j+2} - a_{j+1}) - \frac{3}{h_j} (a_{j+1} - a_j) \\ j = 0, \dots, m-2$$

These are $m-1$ equations for $m+1$ unknowns c_0, c_1, \dots, c_m .

If we use for simplicity natural boundary conditions:

$$S''(x_0) = 2c_0 = 0; \quad S''(x_m) = 2c_m = 0$$

→ get additional eq: $\boxed{c_0 = c_m = 0}$.

Algorithm:

- Input $x_0, \dots, x_m; y_0, \dots, y_m$.
- Build matrix A (see below) and rhs.
- Solve linear system $Ax = b$, $x = \begin{bmatrix} c_0 \\ c_1 \\ \vdots \\ c_m \end{bmatrix}$
- a_j are given by y_j
- b_j ————— (3)
- d_j ————— (**')

recall curvature = $\frac{|f''(x)|}{(1+(f'(x))^2)^{3/2}} \sim |f''(x)|$

spline = curve with minimal curvature passing through knots.

We can now prove Theorem:

Proof: Let $g = f - S$ then $g(x_i) = 0, i = 0, 1, \dots, n$.

$$\int_a^b (f'')^2 dx = \int_a^b (g'')^2 dx + \int_a^b (S'')^2 dx + 2 \int_a^b g'' S'' dx.$$

all we need to show is that $\int_a^b g'' S'' \geq 0$.

$$\int_a^b g'' S'' = \sum_{i=1}^n \int_{x_{i-1}}^{x_i} S'' g'' dx \quad (\text{chopping integral})$$

$$\stackrel{\text{IBP}}{=} \sum_{i=1}^n \left\{ S'' g' \Big|_{x_{i-1}}^{x_i} - \int_{x_{i-1}}^{x_i} S''' g' dx \right\}$$

telescoping sum

$$\stackrel{?}{=} S''(x_n) g'(x_n) - S''(x_0) g'(x_0)$$

$$- \sum_{i=1}^n \int_{x_{i-1}}^{x_i} S''' g' dx \quad \leftarrow \text{constant since } S = \text{cubic}$$

because of natural bdy conditions

$$\stackrel{\text{because of natural bdy conditions}}{=} 0 - \sum_{i=1}^n c_i \int_{x_{i-1}}^{x_i} g' dx$$

$$= - \sum_{i=1}^n c_i \underbrace{(g(x_i) - g(x_{i-1}))}_{=0}$$

$$= 0.$$

QED

CHAPTER 4 NUMERICAL DIFFERENTIATION
AND INTEGRATION

Idea: if we only have pointwise values of a function, we can approximate its derivative (and integral) by using the derivative (or integral) of its polynomial interpolant. Of course if polynomial interpolant is not a good approx between nodes, then our approx derivative (or integral) will not be very good.

§ 4.1 NUMERICAL DIFFERENTIATION

Recall Taylor's theorem:

$$f(x+h) = f(x) + hf'(x) + \frac{h^2}{2} f''(\xi)$$

where ξ is between x and $x+h$.

This gives us a way of approximating derivative since:

(*) $f'(x) = \frac{f(x+h) - f(x)}{h} - \frac{h}{2} f''(\xi)$

- Forward difference : if $h > 0$
- Backward ——— : if $h < 0$

error should decay as $h \rightarrow 0$

the function better be differentiable two times!

Formula (*) was obtained by truncating Taylor's expansion; the truncation error is the term $-\frac{h}{2} f''(\xi)$

A better formula is given by looking at Taylor's theorem: (84)

$$\begin{aligned} (1) \quad f(x+h) &= f(x) + hf'(x) + \frac{h^2}{2} f''(x) + \frac{h^3}{3!} f^{(3)}(\xi_1) \\ (2) \quad f(x-h) &= f(x) - hf'(x) + \frac{h^2}{2} f''(x) - \frac{h^3}{3!} f^{(3)}(\xi_2) \end{aligned}$$

$$(1)-(2): \quad f'(x) = \frac{1}{2h} [f(x+h) - f(x-h)] - \frac{h^2}{12} [f^{(3)}(\xi_1) + f^{(3)}(\xi_2)]$$

Better behaved error term but requires more smoothness of f .

To simplify the error term assume $f^{(3)}$ is continuous over $[x-h, x+h]$

$$\min_{t \in [x-h, x+h]} f^{(3)}(t) \leq c = \frac{f^{(3)}(\xi_1) + f^{(3)}(\xi_2)}{2} \leq \max_{t \in [x-h, x+h]} f^{(3)}(t)$$

Therefore by intermediate value theorem, there is some $\xi \in [x-h, x+h]$ s.t.

$$f^{(3)}(\xi) = c.$$

Thus (22) becomes:

$$f'(x) = \frac{1}{2h} [f(x+h) - f(x-h)] - \frac{h^2}{6} f^{(3)}(\xi).$$

Another useful formula can be obtained by (1)+(2) and adding $\frac{h^4}{4!}$ term. In this case:

$$(***) \quad f''(x) = \frac{1}{h^2} [f(x+h) - 2f(x) + f(x-h)] - \frac{h^2}{12} f^{(4)}(\xi)$$

where $\xi \in (x-h, x+h)$.

We have used same trick as before to simplify error term

(*) (**) and (***) are probably the most used differentiation formulas, and we will use them repeatedly when dealing with ODEs.