

Numerical Analysis

Raphael Hauser
with thanks to Endre Süli

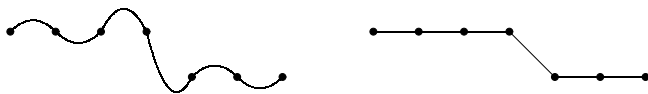
Oxford Mathematical Institute

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Recap on Splines

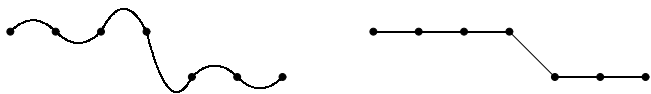


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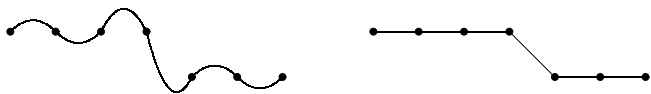
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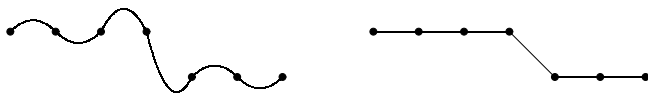
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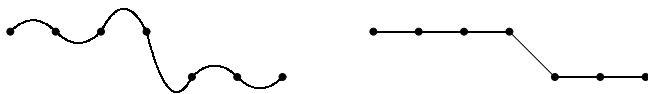
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- C^0 function (linear splines, shown here),
- C^1 function (quadratic splines, not discussed further in this course),
- C^2 function (cubic splines, see below),
- C^3 ... (higher order splines, not discussed in this course).

Theorem

Let $s(x)$ be the linear spline interpolation of a function $f \in C^2[a, b]$ at nodes $x_0 < x_1 < \cdots < x_n$. Then,

$$\|f - s\|_\infty \leq \frac{1}{8}h^2 \|f''\|_\infty$$

where $h = \max_{1 \leq i \leq n} (x_i - x_{i-1})$ and $\|f''\|_\infty = \max_{x \in [a, b]} |f''(x)|$.

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$$s(x) = \begin{cases} \sum_{i=1}^n s_i(x) & \text{for } x \in [a, b] \setminus \{x_0, \dots, x_n\}, \\ f(x_i) & \text{for } x = x_i, \quad (i = 0, \dots, n). \end{cases}$$

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(A) $s(x)$ must interpolate f at the nodes x_0, \dots, x_n ,

$$s_i(x_{i-1}^+) := \lim_{x \searrow x_{i-1}} s_i(x) = f(x_i), \quad (i = 1, \dots, n)$$

$$s_i(x_i^-) := \lim_{x \nearrow x_i} s_i(x) = f(x_{i+1}), \quad (i = 1, \dots, n),$$

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(B) first derivatives of adjacent patches must match at nodes, giving first order smoothness,

$$s'_i(x_i^-) - s'_{i+1}(x_i^+) = 0, \quad (i = 1, \dots, n - 1),$$

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(C) second derivatives of adjacent patches must match at nodes, giving second order smoothness,

$$s''_i(x_i^-) - s''_{i+1}(x_i^+) = 0, \quad (i = 1, \dots, n-1),$$

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- (b) specify $s''(x_0^+) = 0 = s''(x_n^-)$ — this gives a **natural** cubic spline; or
- (c) enforce continuity of s''' at x_1 and x_{n-1} (which implies that the first two pieces are the same cubic spline, i.e., on $[x_0, x_2]$, and similarly for the last two pieces, i.e., on $[x_{n-2}, x_n]$, from which it follows that x_1 and x_{n-1} are not knots! — this is usually described as the '**not a knot**' end-conditions).

We may write Conditions (A), (B) and (C) complemented with one of (a), (b) or (c) in matrix form as

$$Ay = g, \tag{14.1}$$

with

$$y = (a_1, b_1, c_1, d_1, a_2, \dots, d_{n-1}, a_n, b_n, c_n, d_n)^T$$

and the various entries of g are $f(x_i)$, $i \in 0, 1, \dots, n$, and $f'(x_0)$, $f'(x_n)$ for (a), and zeros for (b).

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So if A is nonsingular, this implies that $y = A^{-1}g$, that is there is a unique set of coefficients $\{a_1, b_1, c_1, d_1, a_2, \dots, d_{n-1}, a_n, b_n, c_n, d_n\}$.

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We now prove that if $Ay = 0$ then $y = 0$, and thus that A is nonsingular for cases (a) and (b) — it is also possible, but more complicated, to show this for case (c).

Theorem

If $f(x_i) = 0$ at the knots x_i , $i = 1, \dots, n$, and $f'(x_0) = 0 = f'(x_n)$ for case (a), then $s(x) = 0$ for all $x \in [x_0, x_n]$.

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Proof. Consider

$$\begin{aligned} \int_{x_0}^{x_n} (s''(x))^2 dx &= \sum_{i=1}^n \int_{x_{i-1}}^{x_i} (s''_i(x))^2 dx \\ &= \sum_{i=1}^n [s'_i(x)s''_i(x)]_{x_{i-1}}^{x_i} - \sum_{i=1}^n \int_{x_{i-1}}^{x_i} s'_i(x)s'''_i(x) dx \end{aligned}$$

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using integration by parts. However,

$$\int_{x_{i-1}}^{x_i} s'_i(x)s'''_i(x) dx = s'''_i(x) \int_{x_{i-1}}^{x_i} s'_i(x) dx = s'''_i(x) [s_i(x)]_{x_{i-1}}^{x_i} = 0$$

since $s'''_i(x)$ is constant on the interval (x_{i-1}, x_i) and $s_i(x_{i-1}) = 0 = s_i(x_i)$.

Thus, matching first and second derivatives at the knots, telescopic cancellation gives

$$\begin{aligned}\int_{x_0}^{x_n} (s''(x))^2 dx &= \sum_{i=1}^n [s'_i(x)s''_i(x)]_{x_{i-1}}^{x_i} \\ &= s'_1(x_1)s''_1(x_1) - s'_1(x_0)s''_1(x_0) \\ &\quad + s'_2(x_2)s''_2(x_2) - s'_2(x_1)s''_2(x_1) + \cdots \\ &\quad + s'_{n-1}(x_{n-1})s''_{n-1}(x_{n-1}) - s'_{n-1}(x_{n-2})s''_{n-1}(x_{n-2}) \\ &\quad + s'_n(x_n)s''_n(x_n) - s'_n(x_{n-1})s''_n(x_{n-1}) \\ &= s'_n(x_n)s''_n(x_n) - s'_1(x_0)s''_1(x_0).\end{aligned}$$

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which implies that $s''_i(x) = 0$ and thus $s_i(x) = c_i x + d_i$. Since however $s(x_{i-1}) = 0 = s(x_i)$, $s(x)$ is identically zero on $[x_0, x_n]$. □

Constructing cubic splines. Note that (14.1) provides a constructive method for finding an interpolating spline, but generally this is not used. Motivated by the next result, it is better to find a good basis.

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Proof. If $p, q \in C^2[a, b]$ then $\alpha p + \beta q \in C^2[a, b]$; also $p, q \in \Pi_3$ implies that $\alpha p + \beta q \in \Pi_3$ for every $\alpha, \beta \in \mathbb{R}$.

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Finally, the natural end-conditions (b) imply that

$$(\alpha p + \beta q)''(x_0) = 0 = (\alpha p + \beta q)''(x_n)$$

whenever p'' and q'' are zero at x_0 and x_n . □

Best spline bases: the **Cardinal splines**, C_i , $i = 0, 1, \dots, n$, defined as the interpolatory natural cubic splines satisfying

$$C_i(x_j) = \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j, \end{cases}$$

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Preferred are the **B-splines** (locally) defined by

$$B_i(x_i) = 1 \quad \text{for } i = 2, 3, \dots, n-2,$$

$B_i(x) \equiv 0$ for $x \notin (x_{i-2}, x_{i+2})$, B_i a cubic spline with knots x_j , $j = 0, 1, \dots, n$, with special definitions for B_0 , B_1 , B_{n-1} and B_n .

Example/construction: Cubic B -spline with knots 0, 1, 2, 3, 4. On $[0, 1]$,

$$B(x) = ax^3$$

for some a in order that B , B' and B'' are continuous at $x = 0$ (recall that $B(x)$ is required to be identically zero for $x < 0$). So

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On $[1, 2]$, since B is a cubic polynomial, using Taylor's Theorem,

$$\begin{aligned} B(x) &= B(1) + B'(1)(x-1) + \frac{B''(1)}{2}(x-1)^2 + \beta(x-1)^3 \\ &= a + 3a(x-1) + 3a(x-1)^2 + \beta(x-1)^3 \end{aligned}$$

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for some β , and since we require $B(2) = 1$, then $\beta = 1 - 7a$. Now, in order to continue, by symmetry, we must have $B'(2) = 0$, i.e.,

$$3a + 6a(x-1)_{x=2} + 3(1-7a)(x-1)_{x=2}^2 = 3 - 12a = 0$$

and hence $a = \frac{1}{4}$.

So

$$B(x) = \begin{cases} 0 & \text{for } x < 0 \\ \frac{1}{4}x^3 & \text{for } x \in [0, 1] \\ -\frac{3}{4}(x-1)^3 + \frac{3}{4}(x-1)^2 + \frac{3}{4}(x-1) + \frac{1}{4} & \text{for } x \in [1, 2] \\ -\frac{3}{4}(3-x)^3 + \frac{3}{4}(3-x)^2 + \frac{3}{4}(3-x) + \frac{1}{4} & \text{for } x \in [2, 3] \\ \frac{1}{4}(4-x)^3 & \text{for } x \in [3, 4] \\ 0 & \text{for } x > 4. \end{cases}$$

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More generally: B -spline on $x_i = a + hi$, where $h = (b - a)/n$.

$$B_i(x) = \begin{cases} 0 & \text{for } x < x_{i-2} \\ \frac{(x - x_{i-2})^3}{4h^3} & \text{for } x \in [x_{i-2}, x_{i-1}] \\ -\frac{3(x - x_{i-1})^3}{4h^3} + \frac{3(x - x_{i-1})^2}{4h^2} + \frac{3(x - x_{i-1})}{4h} + \frac{1}{4} & \text{for } x \in [x_{i-1}, x_i] \\ -\frac{3(x_{i+1} - x)^3}{4h^3} + \frac{3(x_{i+1} - x)^2}{4h^2} + \frac{3(x_{i+1} - x)}{4h} + \frac{1}{4} & \text{for } x \in [x_i, x_{i+1}] \\ \frac{(x_{i+2} - x)^3}{4h^3} & \text{for } x \in [x_{i+1}, x_{i+2}] \\ 0 & \text{for } x > x_{i+2}. \end{cases}$$