Module 6 - Spline Interpolation

Lesson goals

- 1. Understanding that splines minimize oscillations by fitting lower-order polynomials to data in a piecewise fashion.
- 2. Recognizing why cubic polynomials are preferable to quadratic and higher-order splines.
- 3. Knowing how to fit a spline to data with MATLAB's built-in functions.

Introduction

There are cases where the polynomial interpolating functions can lead to erroneous results because of round off error and oscillations. In fact, high-degree polynomials can oscillate erratically, that is, a minor fluctuation over a small portion of the interval can induce large fluctuations over the entire range.

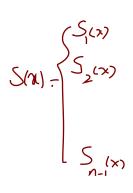
An alternative approach is to apply lower-order polynomials in a piecewise fashion to subsets of data points. Such connecting polynomials are called *spline functions*. For example, third-order curves employed to connect each pair of data points are called *cubic splines*. The key idea in constructing these functions is that the connections between adjacent cubic equations are **visually smooth**.

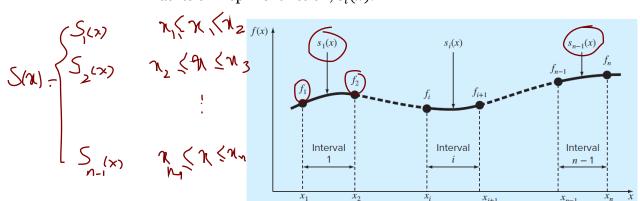
Linear Spline

For *n* data points,

$$(x_1, f_1), (x_2, f_2), ..., (x_n, f_n)$$

there are n-1 subintervals $[x_i, x_{i+1}]$, for i=1,2,...,n. Each i-th subinterval $[x_i, x_{i+1}]$ has its own spline function, $s_i(x)$.





The simplest spline approximation is piecewise-linear interpolation, is merely the straight line connecting the two points at each end of the interval, which is formulated

$$s_i(x) = a_i + b_i(x - x_i)$$

where a_i is the y-intercept, which is defined as

 $a_i = f_i$ and b_i is the slope of the straight line connecting the points:



$$b_{i} = \frac{f_{i+1} - f_{i}}{x_{i+1} - x_{i}}$$

$$S_{i}(x_{1}) = f_{1}$$

$$S_{i}(x_{2}) = f_{2}$$

Example. Fit the data in the following table with first-order splines. Evaluate the Function at x = 5.

S ₁ (x) S ₂ (x) S ₃ (x) 3 4.5 F.0 q.0	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	$S_{1.5} S_{1}(x) = a_{1} + b_{1}(x - 3)$ $S_{2.5} S_{2}(x) = a_{2} + b_{2}(x - 4.5)$ $S_{3}(x) = a_{3} + b_{3}(x - 7)$
$S_{1}(3) = f_{1} = 2.5 \Box \alpha_{1} = 2.5$ $S_{1}(4.5) = f_{2} = 1 \Box \alpha_{1} + b_{1}(4.5)$ $S_{2}(4.5) = 1 \Box \alpha_{2} = 1$ $S_{2}(7) = 2.5 \Box \alpha_{2} + b_{2}(1.5)$	(-3) = -b (b = -1)	5 ₁ 5 ₂ 5 ₃
$\begin{cases} S_3(7) = 0.5 \\ S_3(9) = 0.5 \end{cases} \longrightarrow Q_3 + b_3$	5	- 3 4.5 7 9
$S(x) = \begin{cases} 2.5 - (x-3) \\ 1 + 0.6(x-1) \\ 2.5 - (x-1) \end{cases}$		$f(s) = S(s) = 1 + 0.6 \times 0.5$ = 1.3

Observation. The primary disadvantage of first-order splines is that they are not *smooth*. In essence, at the data points where two splines meet (called a *knot*), the slope changes abruptly. In formal terms, the *first derivative of the function is discontinuous* at these points. This deficiency is overcome by using higher-order polynomial splines that ensure smoothness at the knots by equating derivatives at these points.

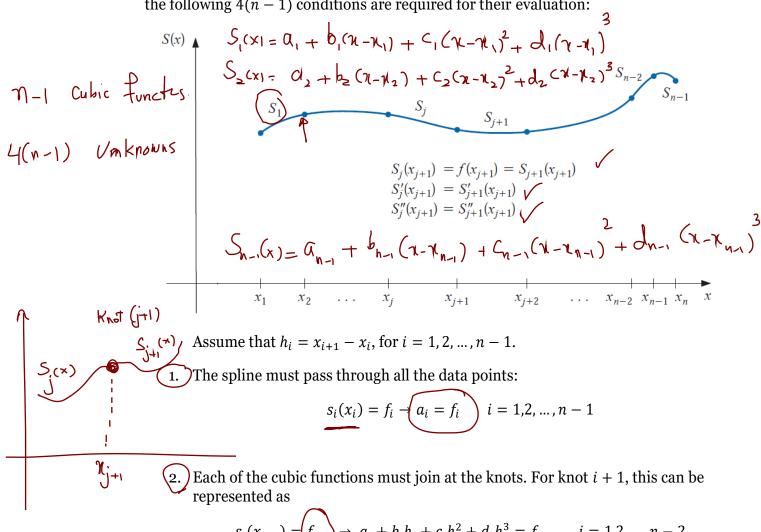
Cubic Splines

The most common piecewise-polynomial approximation uses cubic polynomials between each successive pair of nodes and is called cubic spline interpolation.

The objective in cubic splines is to derive a third-order polynomial for each interval between knots as represented generally by

$$s_i(x) = a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3$$

There are n-1 intervals and 4(n-1) unknown coefficients to evaluate. Consequently, the following 4(n-1) conditions are required for their evaluation:



$$\underbrace{s_i(x_{i+1}) = f_{i+1}}_{s_i(x_{i+1}) = f_{i+1}} \rightarrow \underbrace{a_i + b_i h_i + c_i h_i^2 + d_i h_i^3 = f_{i+1}}_{i+1} \qquad i = 1, 2, \dots, n-2$$
where $h_i = x_{i+1} - x_i$.

3. The first derivatives at the interior nodes must be equal. That is,

$$s'_{i}(x_{i+1}) = s'_{i+1}(x_{i+1}) \rightarrow b_{i} + 2c_{i}h_{i} + 3d_{i}h_{i}^{2} = b_{i+1}$$
 $i = 1, 2, ..., n-2$

4. The second derivatives at the interior nodes must be equal. That is,

$$S_{i}''(x_{i+1}) = S_{i+1}''(x_{i+1}) \to c_i + 3d_ih_i = c_{i+1} \qquad i = 1, 2, ..., n-2$$

After doing multiple substitutions, we can come up with a system of equations in terms of c_i 's (as below) which further enables us to compute the other coefficients by backward substitution.

$$h_{-3} = q.s.$$
Wife $n-1$ Unknown backward substitution.
$$h_{i-1}c_{i-1} + 2(h_{i-1} + h_i)c_i + h_ic_{i+1} = 3(f[x_{i+1}, x_i] - f[x_i, x_{i-1}]) \quad i = 2,3, ..., n-2$$
This system has $n-3$ simultaneous tridiagonal equations with $n-1$ unknown

This system has n-3 simultaneous tridiagonal equations with n-1 unknown coefficients, c_1, c_2, \dots, c_{n-1} . Therefore, if we have two additional conditions, we can solve for the unique *c*'s.

Boundary conditions: Usually, we assume that one of the following sets of

- boundary conditions is satisfied:

 $s''(x_1) = s''(x_n) = 0$ (natural (or *free*) boundary); When the free boundary conditions occur, the spline is called a *natural spline*.
 - • $\sqrt{s'(x_1)} = f_1'$ and $s'(x_n) = f_n'$ (Clamped boundary);

Now, using the natural boundary, the system of equations for unique solution c can be represented as below:

$$\begin{bmatrix} 1 \\ h_1 & 2(h_1 + h_2) & h_2 \\ & & \\ h_{n-2} & 2(h_{n-2} + h_{n-1}) & h_{n-1} \\ & & \\ 1 \end{bmatrix} \begin{bmatrix} c_1 \\ c_2 \\ \vdots \\ c_{n-1} \\ c_n \end{bmatrix} = \begin{bmatrix} 0 \\ 3(f[x_3, x_2] - f[x_2, x_1]) \\ \vdots \\ 3(f[x_n, x_{n-1}] - f[x_{n-1}, x_{n-2}]) \\ 0 \end{bmatrix}$$

Example. Construct a natural cubic spline that passes through the points (1, 2), (2, 3),

$$\begin{cases} S_{1}(x) = Q_{1} + b_{1}(x - 1) + C_{1}(x - 1)^{2} + d_{1}(x - 1) \\ S_{2}(x) = Q_{1} + b_{2}(x - 2) + C_{2}(x - 2)^{2} + d_{2}(x - 2)^{3} \end{cases}$$

$$S_2(x) = Q_2 + b_2(\chi - 2) + C_2(\chi - 2) + d_2(\chi - 2)$$

$$S_{1}^{\prime}(x) = b_{1} + 2C_{1}(x-1) + 3d_{1}(x-1)^{2}$$

$$S_{1}^{\prime}(x) = 2C_{1} + 6d_{1}(x-1)$$

$$S_{1}^{\prime}(x) = 2C_{2} + 6d_{2}(x-2)$$

$$\int_{S_{i}(1)=2}^{\infty} \Rightarrow \boxed{\alpha_{i}=2}$$

$$S_{2}(2) = 3 \Rightarrow 0_{1} = 3$$

$$S_{2}(3) = 5$$

$$G_{2} + b_{2} + c_{1} + c_{2}$$

$$= 5$$

$$9$$
 $S_{2}(3) = 5$
 $G_{2} + G_{2} + G_{2}$
 $G_{2} + G_{2} + G_{2}$

$$\sum_{i=1}^{n} S_{i}(2) = S_{i}(2) \implies Q_{i} + b_{i} + C_{i} + d_{i} = Q_{i} \implies b_{i} + C_{i} + d_{i} = +1$$

3)
$$S_{1}(2) = S_{2}(2)$$
 $\Rightarrow U_{1}+U_{1}$ $\Rightarrow U_{1}+U_{2}$ $\Rightarrow U$

S'(1) = 0 =
$$1$$
 (2(, = 0) - 1 (1) = 0 = 1 (2) 1 (2) 1 (3) = 0 = 1 (2) 1 (2) 1 (2) 1 (3) = 1 (2) 1 (3) = 1 (2) 1 (3) = 1 (3) = 1 (3) = 1 (3) = 1 (3) = 1 (3) = 1 (4) = 1 (4) = 1 (5) = 1 (6) = 1 (7) = 1 (7) = 1 (8) = 1 (8) = 1 (1) = 1

$$S'(1) = 0 \Rightarrow 2C_1 = 0$$

$$S'(1) = 0 \Rightarrow 2C_1 = 0$$

$$S'(3) = 0 \Rightarrow 2C_2 + 6d_2 = 0$$

$$2C_2 + 6d_2 = 0$$

$$2C_3 + 6d_2 = 0$$

$$2C_4 + 6d_2 = 0$$

$$2C_5 + 6d_2 = 0$$

$$2C_6 + 6d_2 = 0$$

$$2C_6 + 6d_2 = 0$$

$$2C_7 + 6d_2 = 0$$

$$b_{2} + c_{2} + d_{2} = 2 \implies 1 + \frac{2}{3}c_{2} + c_{2} - \frac{1}{3}c_{2} = 2$$

$$6 \implies c_{2} = \frac{3}{4}$$

$$5 \quad b_{1} = \frac{3}{4}$$

$$b_{2} = \frac{3}{2}$$

$$d_{3} = -\frac{1}{4}$$

$$S(x) = \begin{cases} 2 + \frac{3}{4}(\chi_{-1}) + \frac{1}{4}(\chi_{-1})^3 & 1 \le \chi \le 2 \\ 3 + \frac{3}{2}(\chi_{-2}) + \frac{3}{4}(\chi_{-2}) - \frac{1}{4}(\chi_{-2})^3 & 2 \le \chi \le 3 \end{cases}$$
Parameters

- 1. Chapra, Steven C. (2018). Numerical Methods with MATLAB for Engineers and Scientists, 4^{th} Ed. McGraw Hill.
- 2. Burden, Richard L., Faires, J. Douglas (2011). *Numerical Analysis*, 9th Ed. Brooks/Cole Cengage Learning