1 Piecewise Cubic Interpolation

Typically the problem with piecewise linear interpolation is the interpolant is not differentiable as the interpolation points (it has a kinks at every interpolation point). A commonly used approach to obtain a smoother interpolation, is to instead replace the linear functions with cubic ones. This means that we will have an interpolation function of the form

$$s(x) = \begin{cases} s_0(x) & \text{if } x_0 \le x < x_1 \\ s_1(x) & \text{if } x_1 \le x < x_2 \\ \vdots & \vdots \\ s_{n-1}(x) & \text{if } x_{n-1} \le x \le x_n \end{cases}$$
(1)

where $s_i(x)$ is a cubic polynomial of the form

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

Since there are n polynomial functions with 4 coefficients each, we have 4n unknowns to solve for. To satisfy the interpolation condition $s(x_i) = y_i$ requires that

$$s_i(x_i) = y_i, \quad i = 0, \dots, n-1, \quad \text{and} \quad s_{n-1}(x_n) = y_n$$
(3)

Note that since $s_i(x_i) = a_i$, the first *n* conditions above immediately imply that $a_i = y_i$ for i = 0, ..., n - 1. Additionally we will impose that s(x) is twice continuously differentiable at the points $x_1, ..., x_{n-1}$. This means

$$s_{i-1}(x_i) = s_i(x_i) = y_i, \quad i = 1, \dots, n-1$$

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

$$s''_{i-1}(x_i) = s''_i(x_i), \quad i = 1, \dots, n-1.$$
(4)

The conditions (3) and (4) are the fundamental conditions that define a cubic spline. Note that in addition to $s_i(x_i) = a_i$ we also have $s'_i(x_i) = b_i$ and $s'_i(x_i) = 2c_i$. If we count up all the constraints, we see that the interpolation condition (3) imposes n + 1 constraints, while twice continuous differentiability (4) imposes 3(n-1) constraints. This amounts to

$$4n - 2 = (n + 1) + 3(n - 1)$$

total constraints, which is obviously 2 less than the 4n coefficients to be determined. Typically this is corrected by imposing two conditions at the end points of the interval. Some of the most common examples are the following

1. Natural Spline: $s_0''(x_1) = s_{n-1}''(x_n) = 0$ This choice produces a spline with the natural property that it minimizes the total curvature of the the approximating spline (this will be discussed in more detail later). 2. Clamped Spline: $s'_0(x_0) = y'_0$ and $s'_{n-1}(x_n) = y'_n$

Here y'_0 and y'_n are either approximations of the derivatives of the function at that point or arbitrarily chosen to pin the slopes at a particular angle.

3. Not-a-Knot Spline: $s_0''(x_1) = s_1''(x_1)$ and $s_{n-2}''(x_{n-1}) = s_{n-1}''(x_{n-1})$

Of course, in order to determine the value of s(x) for any one of these choices, one must solve a complicated linear system for the coefficients a_i, b_i, c_i, d_i , for i = 0, ..., n - 1. As it turns out, such a linear system can be written as a tridiagonal linear system and therefore solved in $\mathcal{O}(n)$ floating point operations.

Example: Lets consider an example. Suppose we want to find the natural cubic spline that interpolates the points (0, 1), (1/2, -1) and (1, 2). The spline take the form

$$s(x) = \begin{cases} s_0(x) & \text{if } 0 \le x \le 1/2\\ s_1(x) & \text{if } 1/2 \le x \le 1, \end{cases}$$
(5)

where

$$s_0(x) = a_0 + b_0 x + c_0 x^2 + d_0 x^3$$

and

$$s_1(x) = a_1 + b_0(x - 1/2) + c_0(x - 1/2)^2 + d_0(x - 1/2)^3$$

From the interpolation conditions (3) and the continuity condition in (4) we have

$$s_0(0) = 1 : \qquad a_0 = 1,$$

$$s_1(1/2) = -1 : \qquad a_1 = -1,$$

$$s_1(1/2) = 2 : \qquad a_1 + \frac{1}{2}b_1 + \frac{1}{4}c_1 + \frac{1}{8}d_1 = 2,$$

$$s_0(1/2) = -1 : \qquad a_0 + \frac{1}{2}b_0 + \frac{1}{4}c_0 + \frac{1}{8}d_0 = -1.$$

Also the derivative conditions from (4) imply that

$$s'_{0}(1/2) = s'_{1}(1/2): \qquad b_{0} + c_{0} + \frac{3}{4}d_{0} = b_{1},$$

$$s''_{0}(1/2) = s''_{1}(1/2): \qquad 2c_{0} + 3d_{0} = 2c_{1}.$$

Finally the condition of being a natural spline implies that

$$s_0''(0) = 0:$$
 $c_0 = 0$
 $s_1''(0) = 0:$ $2c_1 + 3d_1 = 0$

Now one simply need to solve the above linear system for the coefficients. After doing this, one obtains

$$s_0(x) = 1 - \frac{13}{2}x + 10x^3$$

and

$$s_1(x) = -1 + (x - 1/2) + 15(x - 1/2)^2 - 10(x - 1/2)^3.$$

Lets see how this approach works in general. We will assume in what follows that the interpolation points $x_0 < x_1 < \ldots < x_n$ (knots) are equally spaced with $h = x_{i+1} - x_i$ being the spacing between points. As remarked already, the first n conditions in (3) imply that $a_i = y_i$. We still need to determine the remaining 3n coefficients b_i, c_i and d_i for $i = 0, \ldots, n-1$. If we include the last condition of (3) with the continuity condition of (4), then we have

$$s_{i-1}(x_i) = y_i, \quad i = 1, \dots n.$$

Using the fact that $a_i = y_i$, this can be written as

$$y_i + b_i h + c_i h^2 + d_i h^3 = y_{i+1}, \quad i = 0, \dots n - 1.$$
 (6)

When considering the derivative conditions in (4) we may use the fact that $s'_i(x_i) = b_i$ and $s''_i(x_i) = 2c_i$ to conclude that the first derivative conditions $s'_{i-1}(x_i) = s'_i(x_i)$, $i = 1, \ldots, n-1$ can be written as

$$b_i + 2c_i h + 3d_i h^2 = b_{i+1} \quad i = 0, \dots n-2 \tag{7}$$

while the second derivative conditions $s''_{i-1}(x_i) = s''_i(x_i), \quad i = 1, \ldots, n-1$ can be written as

$$c_i + 3d_i h = c_{i+1}$$
 $i = 0, \dots n-2.$ (8)

If we consider the natural cubic spline condition, then we also have $c_0 = \frac{1}{2}s''_0(x_0) = 0$ and $c_{n-1} = \frac{1}{2}s''_{n-1}(x_n) = 0$. In this case it will make the most sense to attempt to reduce solving the linear systems (6), (7), (8) to solving for the coefficients c_i , $i = 1, \ldots, n-2$, first (This is because the natural cubic spline condition are on c_0 and c_{n-1} . If we were to consider clamped splines, then it would make more sense to solve for b_i first.).

To begin, we solve for d_i in equation (8)

$$d_i = \frac{c_{i+1} - c_i}{3h}.$$
 (9)

We may also solve for b_i in (6), which after using the above expression for d_i gives

$$b_{i} = f[x_{i}, x_{i+1}] - c_{i}h - d_{i}h^{2} = f[x_{i}, x_{i+1}] - \frac{2}{3}c_{i}h - \frac{1}{3}c_{i+1}h$$
(10)

where we have used the divided difference notation $f[x_i, x_{i+1}] = (y_{i+1} - y_i)/h$. Finally we can substitute the expression for b_i and the expression for d_i into equation (7) to obtain

$$f[x_i, x_{i+1}] - \frac{2}{3}c_ih - \frac{1}{3}c_{i+1}h + 2c_ih + (c_{i+1} - c_i)h = f[x_{i+1}, x_{i+2}] - \frac{2}{3}c_{i+1}h - \frac{1}{3}c_{i+2}h$$

Rearranging this and collecting terms gives an equation for c_i , i = 1, ..., n - 2

$$\frac{1}{3}c_i + \frac{4}{3}c_{i+1} + \frac{1}{3}c_{i+2} = \frac{f[x_{i+1}, x_{i+2}] - f[x_i, x_{i+1}]}{h} = f[x_i, x_{i+1}, x_{i+2}].$$

Finally, using the fact that for the natural spline $c_0 = 0$ and $c_{n-1} = 0$ gives the following linear system

$$\frac{1}{3} \begin{bmatrix}
4 & 1 & 0 & 0 & \cdots & 0 \\
1 & 4 & 1 & 0 & \cdots & 0 \\
0 & 1 & 4 & 1 & \ddots & \vdots \\
\vdots & \ddots & \ddots & \ddots & \ddots & \vdots \\
0 & 0 & 0 & 1 & 4 & 1 \\
0 & 0 & 0 & 0 & 1 & 4
\end{bmatrix} \begin{bmatrix}
c_1 \\
c_2 \\
\vdots \\
\vdots \\
c_{n-3} \\
c_{n-2}
\end{bmatrix} = \begin{bmatrix}
f[x_0, x_1, x_2] \\
f[x_1, x_2, x_3] \\
\vdots \\
f[x_{n-3}, x_{n-2}, x_{n-1}] \\
f[x_{n-2}, x_{n-1}, x_n]
\end{bmatrix}$$
(11)

Note that this is a tridiagonal linear system and is strictly diagonally dominant since 4 > 1+1. Therefore the matrix is positive definite and non-singular and can be solved quickly using the Thomas algorithm with $\mathcal{O}(n)$ flops. Once you solve this system, for c_i , $i = 1, \ldots n - 1$ the other coefficients b_i and d_i can be obtained from equations (10) and (9), each costs $\mathcal{O}(n)$ flops.

1.1 Cubic B-Splines

As another approach to the computation of cubic splines, we will find useful follow an approach similar to piecewise linear interpolation or Lagrange interpolation approach, introducin a family of basis functions. Our goal is to write the spline as

$$s(x) = \sum_{i} a_i B_i(x)$$

where $B_i(x)$ is a certain piecewise cubic spline. The functions $B_i(x)$ are called *cubic B-Splines*. Although similar to the formulas for piecewise linear interpolation and Lagrange interpolation formulas, there is an important difference. The coefficients a_i are not necessarily equal to the function output values y_i .

The main benefit of this approach is that the interpolation procedure can be broken into two pieces. First construct the B-splines $B_i(x)$ and then find the a_i 's. The has the benefit that, since the construction of the B-splines will not in any way depend on the y_i data, meaning, once you have computed the B-splines you do not need to compute them again for different sets of y_i values.

Computing the B-splines

We will assume that each $B_i(x)$ is piecewise cubic function of the following form

$$B_{i}(x) = \begin{cases} 0 & \text{if } x \leq x_{i-2}, \\ q_{i-2}(x) & \text{if } x_{i-2} \leq x_{i-1}, \\ q_{i-1}(x) & \text{if } x_{i-1} \leq x_{i}, \\ q_{i+1}(x) & \text{if } x_{i} \leq x_{i+1}, \\ q_{i+2}(x) & \text{if } x_{i+1} \leq x_{i+2}, \\ 0 & \text{if } x_{i+2} \leq x, \end{cases}$$

where $q_i(x)$, i = 0, ..., n are the usual cubic polynomials of the form

$$q_i(x) = A_i + B_i(x - x_i) + C_i(x - x_i)^2 + D_i(x - x_i)^3.$$

The functions q_i will be determined by the requirement that

$$B_i(x_{i-2}) = B'_i(x_{i-2}) = B''_i(x_{i-2}) = 0$$
(12)

and

$$B_i(x_{i+2}) = B'_i(x_{i+2}) = B''_i(x_{i+2}) = 0$$

and that $B_i(x)$ is twice continuously differentiable at x_{i-1}, x_i, x_{i+1} . Since we are assuming the interpolation points are equally spaced, this will imply that $B_i(x)$ is symmetric about $x = x_i$. Note that the two continuous derivatives and the symmetry automatically imply that $B'_i(x_i) = 0$. Using symmetry, it follows that we only need to ensure equation (12) holds, that the two continuous derivatives at x_{i-1} match and that $B'_i(x_i) = 0$. This is accomplished by requiring the following

$$q_{i-2}(x_{i-2}) = q'_{i-2}(x_{i-2}) = q''_{i-2}(x_{i-2}) = 0$$
$$q_{i-2}(x_{i-1}) = q_{i-1}(x_{i-1}), \quad q'_{i-2}(x_{i-1}) = q'_{i-1}(x_{i-1}), \quad q''_{i-2}(x_{i-1}) = q''_{i-1}(x_{i-1})$$

as well as

$$q_{i-1}'(x_i) = 0.$$

From the conditions at x_{i-2} we can easily conclude (by integration) that

$$q_{i-2} = D_{i-2}(x - x_{i-2})^3.$$

Additionally, from the remaining smoothness conditions at x_{i-1} and the zero derivative condition at x_i one finds that

$$q_{i-1}(x) = D_{i-2} \left(h^3 + 3h^2 (x - x_{i-1}) + 3h(x - x_{i-1})^2 - 3(x - x_{i-1})^3 \right)$$

Of course, since we needed to determine 8 coefficients for q_{i-2} and q_{i-1} with only 7 constraints, we have an undetermined constant D_{i-1} . Typically the convention is to take $B_i(x_i) = 2/3$ which sets $D_i = 1/6h^3$. Using the symmetry of $B_i(x)$ about $x = x_i$ we automatically obtain $q_{i+1}(x)$ and $q_{i+2}(x)$ by the formulas

$$q_{i+1}(x) = q_{i-1}(2x_i - x), \quad q_{i+2}(x) = q_{i-2}(2x_i - x).$$

After doing this, we find that $B_i(x)$ is completely determined by

$$q_{i-2} = \frac{1}{6h^3}(x - x_{i-2})^3$$

$$q_{i-1} = \frac{1}{6} + \frac{1}{2h}(x - x_{i-1}) + \frac{1}{2h^2}(x - x_{i-1})^2 - \frac{1}{2h^3}(x - x_{i-1})^3$$

$$q_{i+1} = \frac{1}{6} - \frac{1}{2h}(x - x_{i+1}) + \frac{1}{2h^2}(x - x_{i+1})^2 - \frac{1}{2h^3}(x - x_{i+1})^3$$

$$q_{i+2} = -\frac{1}{6h^3}(x - x_{i+2})^3$$

By factoring the polynomials in the above formula, it is possible to show that

$$B_i(x) = B\left(\frac{x - x_i}{h}\right)$$

where

$$B(x) = \begin{cases} \frac{2}{3} - x^2 \left(1 - \frac{1}{2}|x|\right) & \text{if } |x| < 1, \\ \frac{1}{6}(2 - |x|)^3 & \text{if } 1 \le |x| \le 2, \\ 0 & \text{if } 2 \le |x|. \end{cases}$$

Finding the coefficients

Now we write our spline as

$$s(x) = \sum_{i=-1}^{n+1} a_i B_i(x)$$

Note that we have taken the sum over $i = -1, \ldots, n+1$ even though we do not have points $x_{-3}, x_{-2}, x_{-1}, x_{n+1}, x_{n+2}, x_{n+3}$. This is because we want to sum over all B_i which are non-zero on the interval $[x_0, x_n]$. When a piece of $B_i(x)$ lies outside of the interval $[x_0, x_n]$, we simply ignore that piece. Any piece of $B_i(x)$ that lies inside $[x_0, x_n]$ will, by the piecewise polynomial nature of $B_i(x)$, necessarily only depend on the points x_i that belong to x_0, \ldots, x_n .

The task of determining the coefficients a_i is now simple once you realize that at any point x_i only $B_{i-1}(x_i)$, $B_i(x_i)$ and $B_{i+1}(x_i)$ are non-zero (i.e. they are the only functions 'turned on' at x_i). Using the fact that for each i = 1, ..., n-1 we have the exact values

$$B_i(x_{i-1}) = \frac{1}{6}, \quad B_i(x_i) = \frac{2}{3}, \quad B_i(x_{i+1}) = \frac{1}{6}$$

We see that

$$s(x_i) = \frac{1}{6} \left(a_{i-1} + 4a_i + a_{i+1} \right)$$

and therefore, the interpolation requirement $s(x_i) = y_i$ for i = 0, ..., n implies that

$$a_{i-1} + 4a_i + a_{i+1} = 6y_i$$
 for $i = 1, 2, \dots, n$.

In order to solve for a_i we need to know the values of a_{-1} and a_{n+1} . Again, this is where the additional endpoint conditions are used. It is not difficult to show that for the natural spline, we have

$$s''(x_i) = a_{i-1}B''_i(x_i) + a_iB''_i(x_i) + a_{i-1}B''_{i+1}(x_i) = \frac{1}{h^2}(a_{i-1} - 2a_i + a_{i+1}).$$

Solving the natural spline conditions $s''(x_0) = 0$ and $s''(x_n) = 0$ using the above formula give

$$a_{-1} = 2a_0 - a_1, \quad a_{n+1} = 2a_n - a_{n-1}.$$

This means when i = 0, we get $a_0 = y_0$ and when i = n, we get $a_n = y_n$. Now the remaining linear system for a_i , i = 1, ..., n - 1 can be solved by a tridiagonal system

Γ4	1	0	0	• • •	0	$\begin{bmatrix} a_1 \end{bmatrix}$		$\begin{bmatrix} 6y_1 - y_0 \end{bmatrix}$
1	4	1	0	• • •	0	a_2		$6y_2$
0	1	4	1	·	:		_	÷
:	۰.	·	·	·	:		_	
0	0	0	1	4	1	a_{n-2}		$6y_{n-2}$
0	0	0	0	1	4	$\lfloor a_{n-1} \rfloor$		$6y_{n-1} - y_n$

•

Again we find that the remaining coefficients can be computed by inverting a tridiagonal, positive definite system. The main difference between this system and the tridiagonal system obtained (11) lies in the right-hand side. In (11) the right-hand side involves computing n-2 2nd order divided differences, while the right-hand side of the above linear system only involves the values y_i , i = 0, ..., n. In some sense we have already done this computation in computing the B-splines $B_i(x)$.