

# Chapter 3: Splines

## Contents

<b>1</b>	<b>Introduction</b>	<b>1</b>
<b>2</b>	<b>Problem Setting</b>	<b>1</b>
<b>3</b>	<b>Examples</b>	<b>2</b>
<b>4</b>	<b>Linear Spline</b>	<b>3</b>
4.1	Accuracy Theorem for linear spline . . . . .	4
<b>5</b>	<b>Natural Cubic Spline</b>	<b>4</b>
<b>6</b>	<b>Smoothness Theorem for natural splines</b>	<b>6</b>

## 1 Introduction

Disdvantages of Polynomial Interpolation  $P_n(x)$ :

- n-time differentiable. We do not need such high smoothness.
- big error in certain intervals (esp near the end points)
- There is no convergence results
- Heavy to compute for large n

Suggestion: use piecewise polynomial interpolation

Usage:

- Visualization of discrete data
- graphic design

Requirement:

- Interpolation
- certain degree of smoothness

## 2 Problem Setting

Given a set of data

$$\begin{array}{c|c|c|c|c} x & t_0 & t_1 & \dots & t_n \\ y & y_0 & y_1 & \dots & y_n \end{array}$$

Find a function  $S(x)$  which interpolates points  $(t_i, y_i)$  for all  $i$ .

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

Figure 1:

The set  $t_0 < t_1 < \dots < t_n$  are called knots. Note they need to be ordered.

$S(x)$  consists of piecewise polynomials.

$S(x)$  is called a spline of degree  $k$ , if

- $S_i(x)$  is a polynomial of degree  $k$
- $S(x)$  is  $(k-1)$  times continuous differentiable i.e. for  $i = 1, 2, \dots, k-1$  we have:

$$S_{i-1}(t_i) = S_i(t_i)$$

$$S'_{i-1}(t_i) = S'_i(t_i)$$

...

$$S^{(k-1)}_{i-1}(t_i) = S^{(k-1)}_i(t_i)$$

Commonly used splines:

- $n = 1$ : linear spline (simplest)
- $n = 2$ : quadratic spline (less popular)
- $n = 3$ : cubic spline (most used)

### 3 Examples

Determine whether this function is a first-degree spline function.

**Answer** Check the properties of a linear spline.

- Linear polynomial for each piece - OK (degree 1 or less)
- $S(x)$  is continuous at inner knots
- At  $x = 0$ ,  $S(x)$  is discontinuous, because from the left we get 0 and from the right we get 1.

Therefore this is NOT linear spline.

Determine whether the following function is a quadratic spline.

**Answer** Let's label each piece:

$$S(x) = \begin{cases} x & x \in [-1, 0] \\ 1 - x & x \in (0, 1) \\ 2x - 2 & x \in [1, 2] \end{cases}$$

Figure 2:

$$S(x) = \begin{cases} x^2 & x \in [-10, 0] \\ -x^2 & x \in (0, 1) \\ 1 - 2x & x \geq 1 \end{cases}$$

Figure 3:

$$Q_0(x) = x^2; Q_1(x) = -x^2; Q_2(x) = 1 - 2x$$

We now check all conditions. Let's check the continuity at the inner knots of Q and Q'.

$$Q_0(0) = 0 = Q_1(0)$$

$$Q_1(1) = -1 = Q_2(1)$$

$$Q'_0(0) = 0 = Q'_1(0)$$

$$Q'_1(1) = -2 = Q'_2(1)$$

Therefore, since all conditions pass, this IS a quadratic spline.

## 4 Linear Spline

n = 1: piecewise linear interpolation, i.e., straight line between 2 neighboring points.

Requirements:

$$S_0(t_0) = y_0$$

$$S_{i-1}(t_i) = S_i(t_i) = y_i; i = 1, 2, \dots, n-1$$

$$S_{n-1}(t_n) = y_n$$

Easy to find: write the equation for a line through two points  $(t_i, y_i)$  and  $(t_{i+1}, y_{i+1})$

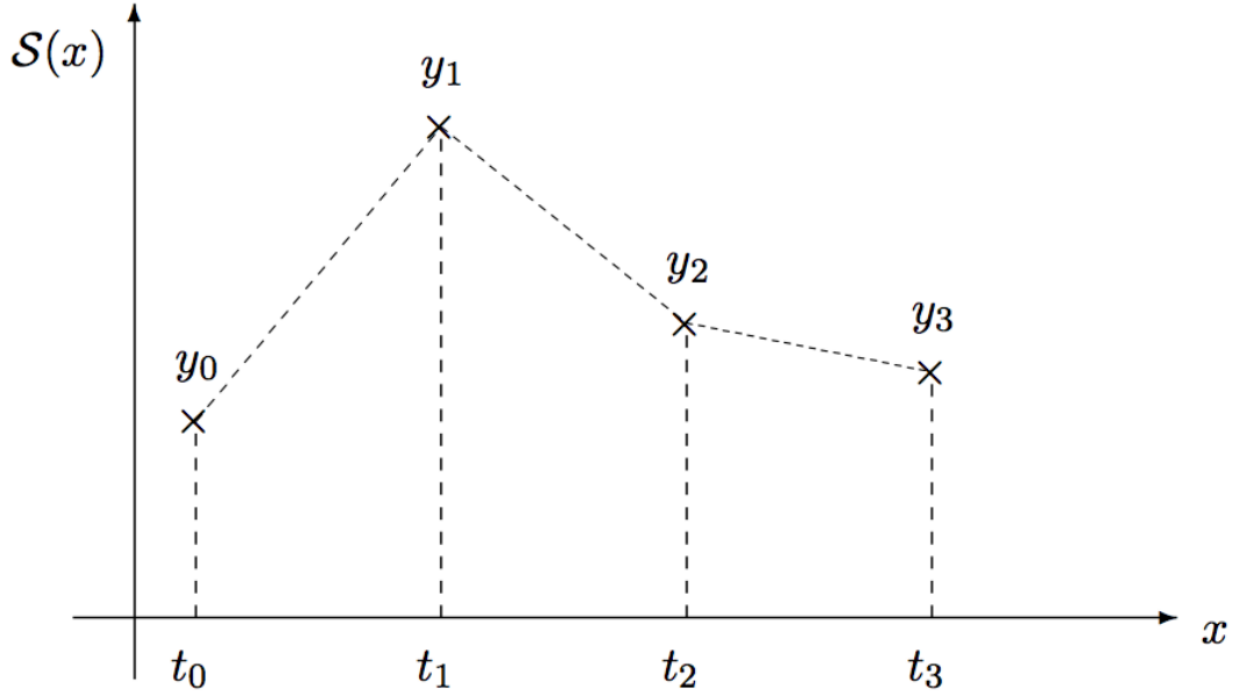


Figure 4:

$$S_i(x) = y_i + \frac{y_{i+1} - y_i}{t_{i+1} - t_i}(x - t_i); i = 0, 1, 2, \dots, n - 1$$

#### 4.1 Accuracy Theorem for linear spline

Assume  $t_0 < t_1 < \dots < t_n$  and let  $h_i = t_{i+1} - t_i, h = \max_i h_i$

$f(x)$  is the given function and  $S(x)$  is the linear spline that interpolates the function such that

$$S(t_i) = f(t_i); i = 0, 1, \dots, n$$

Then we have the following for  $x \in [t_0, t_n]$

- If  $f''$  exists and is continuous, then  $|f(x) - S(x)| \leq \frac{1}{8}h^2 \max_x |f''(x)|$
- If  $f'$  exists and is continuous, then  $|f(x) - S(x)| \leq \frac{1}{2}h \max_x |f'(x)|$

To minimize error, it is obvious that one should add more knots where the function has large first or second derivative.

Since a quadratic spline is less used, please reference it within the book.

## 5 Natural Cubic Spline

Given  $t_0 < t_1 < \dots < t_n$ , we define the cubic spline with

$$S(x) = S_i(x); t_i \leq x \leq t_{i+1}$$

Write

$$S_i(x) = a_i x^3 + b_i x^2 + c_i x + d_i, i = 0, 1, \dots, n-1$$

Total number of unknowns is 4 \* n. Requirements are that S, S', S'' are all continuous.

Equations we have:

- $S_i(t_i) = y_i, i = 0, 1, \dots, n-1$
- $S_i(t_{i+1}) = y_{i+1}, i = 0, 1, \dots, n-1$
- $S'_i(t_{i+1}) = S'_{i+1}(t_{i+1}), i = 0, 1, \dots, n-2$
- $S''_i(t_{i+1}) = S''_{i+1}(t_{i+1}), i = 0, 1, \dots, n-2$

This gives us 4n-2 equations total, so we still need two more conditions.

- $S''_0(t_0) = 0, S''_{n-1}(t_n) = 0$

These last two equations choice makes our spline a “natural” spline.

How to compute  $S_i(x)$ ? We know:

- $S_i$  is polynomial degree 3
- $S'_i$  is degree 2
- $S''_i$  is degree 1

Procedure

1. Start with  $S''_i(x)$ , they are all linear, one can use Lagrange form
2. Integrate  $S''_i(x)$  twice to get  $S_i(x)$ , you will get 2 integration constant
3. Determine these constants. Various tricks on the way...

Define:  $z_i = S''(t_i), i = 1, 2, \dots, n-1, z_0 = z_n = 0$

These  $z_i$ 's are our unknowns.

Let  $h_i = t_{i+1} - t_i$ . Lagrange form for  $S''_i$ .

$$S''_i(x) = \frac{z_{i+1}}{h_i}(x - t_i) - \frac{z_i}{h_i}(x - t_{i+1})$$

Then

$$S'_i(x) = \frac{z_{i+1}}{2h_i}(x - t_i)^2 - \frac{z_i}{2h_i}(x - t_{i+1})^2 + C_i - D_i$$

$$S_i(x) = \frac{z_{i+1}}{6h_i}(x - t_i)^3 - \frac{z_i}{6h_i}(x - t_{i+1})^3 + c_i(x - t_i) - D_i(x - t_{i+1})$$

You can check yourself that these two equations are correct.

Interpolating Properties:  $S_i(t_i) = y_i$

$$\begin{cases} h_{i-1}z_{i-1} + 2(h_{i-1} + h_i)z_i + h_i z_{i+1} = 6(b_i - b_{i-1}), & i = 1, 2, \dots, n-1 \\ z_0 = z_n = 0. \end{cases}$$

Figure 5:

$$y_i = -\frac{z_i}{6h_i}(-h_i)^3 - D_i(-h_i) = \frac{1}{6}z_i h_i^2 + D_i h_i$$

$$D_i = \frac{y_i}{h_i} - \frac{h_i}{6}z_i$$

$S_i(t_{i+1}) = y_{i+1}$  gives

$$y_{i+1} = \frac{z_{i+1}}{6h_i}h_i^3 + C_i h_i$$

$$C_i = \frac{y_{i+1}}{h_i} - \frac{h_i}{6}z_{i+1}$$

We see that, once  $z_i$ 's are known, then  $(C_i, D_i)$ 's are known, and so are  $S_i, S_i'$  are known.

$$S_i(x) = \frac{z_{i+1}}{6h_i}(x - t_i)^3 - \frac{z_i}{6h_i}(x - t_{i+1})^3 + \left(\frac{y_{i+1}}{h_i} - \frac{h_i}{6}z_{i+1}\right)(x - t_i) - \left(\frac{y_i}{h_i} - \frac{h_i}{6}z_i\right)(x - t_{i+1})$$

$$S_i'(x) = \frac{z_{i+1}}{2h_i}(x - t_i)^2 - \frac{z_i}{2h_i}(x - t_{i+1})^2 + \frac{y_{i+1} - y_i}{h_i} - \frac{z_{i+1} - z_i}{6}h_i$$

Continuity of  $S'(x)$  requires

$$S'_{i-1}(t_i) = S'_i(t_i), i = 1, 2, \dots, n-1$$

$$S'_i(t_i) = -\frac{1}{6}h_i z_{i+1} - \frac{1}{3}h_i z_i + b_i$$

$$S'_{i-1}(t_i) = \frac{1}{6}z_{i-1}h_{i-1} + \frac{1}{3}z_i h_{i-1} + b_{i-1}$$

Set them equal to each other and we get:

Summarizing the algorithm:

- Set up the matrix-vector equation and solve for  $z_i$
- Compute  $S_i(x)$  using these  $z_i$ 's

See Matlab codes.

## 6 Smoothness Theorem for natural splines

$$\begin{cases} h_{i-1}z_{i-1} + 2(h_{i-1} + h_i)z_i + h_i z_{i+1} = 6(b_i - b_{i-1}), & i = 1, 2, \dots, n-1 \\ z_0 = z_n = 0. \end{cases}$$

In matrix-vector form:  $\mathbf{H} \cdot \vec{z} = \vec{b}$

$$\mathbf{H} = \begin{pmatrix} 2(h_0 + h_1) & h_1 & & & \\ h_1 & 2(h_1 + h_2) & h_2 & & \\ & h_2 & 2(h_2 + h_3) & h_3 & \\ & & \ddots & \ddots & \ddots \\ & & & h_{n-3} & 2(h_{n-3} + h_{n-2}) & h_{n-2} \\ & & & & h_{n-2} & 2(h_{n-2} + h_{n-1}) \end{pmatrix}$$

$\mathbf{H}$ : tri-diagonal, symmetric, and diagonal dominant

$$2|h_{i-1} + h_i| > |h_i| + |h_{i-1}|$$

which implies unique solution.

Figure 6:

$$\begin{cases} h_{i-1}z_{i-1} + 2(h_{i-1} + h_i)z_i + h_i z_{i+1} = 6(b_i - b_{i-1}), & i = 1, 2, \dots, n-1 \\ z_0 = z_n = 0. \end{cases}$$

$$\vec{z} = \begin{pmatrix} z_1 \\ z_2 \\ z_3 \\ \vdots \\ z_{n-2} \\ z_{n-1} \end{pmatrix}, \quad \vec{b} = \begin{pmatrix} 6(b_1 - b_0) \\ 6(b_2 - b_1) \\ 6(b_3 - b_2) \\ \vdots \\ 6(b_{n-2} - b_{n-3}) \\ 6(b_{n-1} - b_{n-2}) \end{pmatrix}.$$

Figure 7:

**Theorem.** If  $S$  is the natural cubic spline function that interpolates a twice-continuously differentiable function  $f$  at knots

$$a = t_0 < t_1 < \cdots < t_n = b$$

then

$$\int_a^b [S''(x)]^2 dx \leq \int_a^b [f''(x)]^2 dx.$$

Note that  $\int (f'')^2$  is related to the curvature of  $f$ .

Cubic spline gives the least curvature,  $\Rightarrow$  most smooth, so best choice.

Figure 8:

**Proof.** Let

$$g(x) = f(x) - S(x)$$

Then

$$g(t_i) = 0, \quad i = 0, 1, \dots, n$$

and

$$f'' = S'' + g'', \quad (f'')^2 = (S'')^2 + (g'')^2 + 2S''g''$$

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

We claim that

$$\int_a^b S''g'' dx = 0$$

then this would imply

$$\int_a^b (f'')^2 dx \geq \int_a^b (S'')^2 dx$$

and we are done.

Figure 9:



Proof of the claim:

$$\int_a^b S'' g'' dx = 0$$

Using integration-by-parts,

$$\int_a^b S'' g'' dx = S'' g' \Big|_a^b - \int_a^b S''' g' dx$$

Since  $S''(a) = S''(b) = 0$ , the first term is 0.

For the second term, since  $S'''$  is piecewise constant. Call

$$c_i = S'''(x), \quad \text{for } x \in [t_i, t_{i+1}].$$

Then

$$\int_a^b S''' g' dx = \sum_{i=0}^{n-1} c_i \int_{t_i}^{t_{i+1}} g'(x) dx = \sum_{i=0}^{n-1} c_i [g(t_{i+1}) - g(t_i)] = 0,$$

(b/c  $g(t_i) = 0$ ).

Figure 10: