CE 601: Numerical Methods

Cubic Splines

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Cubic Splines

- Till now, we wee dealing with polynomials that approximate the actual function for a given data set.
- For a given (n+1) dat set, we were able to develop <u>exact-fit</u> polynomials.

 $P_1(x) \rightarrow$ if only two of the (n+1) data used

 $P_2(x) \rightarrow$ if only three of the (n+1) data used, etc.

 $P_n(x) \rightarrow$ the unique exactly-fit polynomial that passes through all the given (n+1) data points.

- So we can see that for given data set, higher the polynomial order – greater the chance the polynomial passes through all the data points.
- Higher degree polynomials, therefore have few errors at the data points.

• What happens if we increase the degree of the polynomial.



- If we use higher degree polynomials it is passing through all the data points (as shown).
- However by looking into the shape, you may be having a feeling that in between data points, the polynomial may not be predicting the value accurately.

- Most of the times for higher degree polynomials this happens.
- This is true for most of the higher order polynomials.
- Therefore two questions remain in front of you:
- What is the point in going for higher degree polynomials?
- The lower degree polynomials will not pass through all the points.
- To overcome such issues, we can use splines.
- What is spline?
- Recall in your engineering graphics days, you used to take splines (or flexible rods) and use them to draw curves between multiple points.



Fig. Spline

(ref: <u>http://en.wikipedia.org/wiki/Flat_spline#mediaviewer/File:Spline_%28PSF%29.png</u>

- The same philosophy is used here.
- You will connect the data points in piece-wise manner.
- The polynomial connecting (x_0, f_0) and (x_1, f_1) may be different from (x_1, f_1) and (x_2, f_2) , etc.
- Splines are lower degree polynomials that connect sub-sets of data points.
- First-degree splines will be straight line between each points.
- While incorporating splines, we have to see that the function is consistent as per actual function (i.e. it should be differentiable, the slopes and curvatures are continuous).

- In linear splines, we miss the continuity of first and second derivatives (i.e. slopes and curvature).
- If we use second degree polynomials, we miss continuity of curvature.
- Therefore cubic splines (third degree spline) are the lower degree polynomials that can assure continuity of slope as well as curvature.
- <u>To develop cubic spline</u>
- Consider the (n+1) data points $(x_0, f_0), (x_1, f_1), \dots, (x_n, f_n)$ in a data line below:



- There are (n+1) data points and *n* intervals.
- At each interval we want to develop cubic splines. Intervals *j* = 1, 2, 3, ..., *n*

Cubic spline for the j^{th} interval is given as:

$$f_{j}(x) = a_{j} + b_{j}x + c_{j}x^{2} + d_{j}x^{3}$$

where for j^{th} interval, if $j = i, x_{i-1} \le x \le x_i$.

$$\therefore \text{ We can write } f_i(x) = a_i + b_i x + c_i x^2 + d_i x^3$$

- The conditions for cubic spline approximations
- (i) Function values at data points are known i.e. $f(x_0) = f_0$, $f(x_n) = f_n$ Also $f(x_i) = f_i$
- → The data point x_i will be part of two splines $f_i(x)$ and $f_{i+1}(x)$. So, we require $f_i(x_i) = f_{i+1}(x_i) = f_i$

(ii) The derivatives or slope should be same:

$$f'_{i}(x) = b_{i} + 2c_{i}x + 3d_{i}x^{2}$$

$$f_{i+1}'(x) = b_{i+1} + 2c_{i+1}x + 3d_{i+1}x^{2}$$

$$f'_{i}(x_{i}) = f_{i+1}'(x_{i})$$

(iii) The second derivative $f_i''(x) = 2c_i + 6d_i x$ We also require $f_i''(x_i) = f_{i+1}''(x_i)$

(iv) The first spline i.e. $f_1(x)$ should pass through $x = x_{0.}$

The last spline i.e. $f_2(x)$ should pass through $x = x_n$.

(v) At $x = x_{0.}$; $f_1''(x_0) = \text{specified } f''(x_0)$ (should be known) At $x = x_{n.}$; $f_n''(x_n) = \text{specified } f''(x_n)$) (should be known)

$$\rightarrow f_i(x) = a_i + b_i x + 2c_i x^2 + 3d_i x^3$$

$$f_i'(x) = b_i + 2c_i x + 3d_i x^2$$

$$f_i''(x) = 2c_i + 6d_i x$$

• This is a linear equation. We can represent it by Lagrange polynomial.

$$f_i''(x) = \frac{x - x_i}{x_{i-1} - x_i} f_i''(x_{i-1}) + \frac{x - x_{i-1}}{x_i - x_{i-1}} f_i''(x_i)$$

As per the cubic spline criteria: $f_i''(x_{i-1}) = f_{i-1}''(x_{i-1}) = f_{i-1}''$ Similarly, $f_i''(x_i) = f_i''$ Also, $f_i'(x_{i-1}) = f_{i-1}'(x_{i-1}) = f_{i-1}'$ and $f_i'(x_i) = f_{i+1}'(x_{i-1}) = f_i'$ $\therefore f_i'(x) = \int f_i''(x_i) dx = \frac{\frac{x^2}{2} - xx_i}{x_{i-1}} f_{i-1}'' + \frac{\frac{x^2}{2} - xx_{i-1}}{x_{i-1}} f_i'' + C$

and

$$f_i(x) = \int f_i'(x_i) dx = \frac{\frac{x^3}{6} - \frac{x^2}{2}x_i}{x_{i-1} - x_i} f_{i-1}'' + \frac{\frac{x^3}{6} - \frac{x^2}{2}x_{i-1}}{x_i - x_{i-1}} f_i'' + Cx + D$$

- To find the integration parameters *C* and *D*, we have to utilize the known values at $x = x_{i-1}$ and $x = x_i$. i.e. $f'_i(x_i) = f'_i, f'_i(x_{i-1}) = f_{i-1}, f_i(x_i) = f_i, f_i(x_{i-1}) = f_{i-1}$
- However in actual, we have only the values $f_0, f_1, f_2, \dots, f_n$ -> (n+1) values f_0' and f_n' -> 2 values f_0'' and f_n''' -> 2 values

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0-	-17 @	Ø	S
1-2	1-1	4	×+1
2.2	× ji-i	\varkappa_i	\mathcal{L}_{i+i}
Si-2	8	8.	Sie.
8' 2	2'	8.	S' and
8"	8"	8."	8"

• All other f'_i and f''_i should be inferred from these values.

So for any interval,

$$\begin{split} f_i(x) &= \frac{(x_i - x)^3}{6(x_i - x_{i-1})} f_{i-1}^{"} + \frac{(x - x_{i-1})^3}{6(x_i - x_{i-1})} f_i^{"} \\ &+ (x_i - x) \left\{ \frac{1}{(x_i - x_{i-1})} f_{i-1} - \frac{(x_i - x_{i-1})}{6} f_{i-1}^{"} \right\} \\ &+ (x - x_{i-1}) \left\{ \frac{f_i}{(x_i - x_{i-1})} - \frac{(x_i - x_{i-1})}{6} f_i^{"} \right\} \to (1) \end{split}$$

• Differentiating this equation and applying at known points f_{i+1} , f_i and f_{i+1} , finally we will get corresponding expressions.

- To find expressions for f_i'' at interval nodes i=1, 2, 3, ..., n-1obtained by differentiating eq. (1) once to get $f_i'(x)$ and again for $f_i''(x)$.
- The differentiated eq. $f'_i(x)$ is used for two known points f_{i-1} and f'_i .
- Also differentiated eq. $f_{i+1}'(x)$ is used for f'_i .
- Similarly, $f_{i-1}'(x)$ is used for f_{i-1}' .
- Equating these equations, we get unknown in terms of f_{i-1}", f_i" and f_{i+1}".

$$\rightarrow (x_{i} - x_{i-1}) f_{i+1} "+ 2(x_{i+1} - x_{i-1}) f_{i} "+ (x_{i+1} - x_{i}) f_{i+1} "$$

$$= 6 \frac{f_{i+1} - f_{i}}{x_{i+1} - x_{i}} - 6 \frac{f_{i} - f_{i-1}}{x_{i} - x_{i-1}} \rightarrow (2)$$

Eq. (2) is substituted in (n-1) interval nodes to get systems of equations for unknown f_i " (i = 1, 2, 3, ..., n-1).

- **Example** (As adopted from the course text book Hoffman's Numerical Methods)
- For the given data set

	J						
i	X	f(x)	f"(x)				
0	-0.500	0.731531	0.0	@ •	(*)	2	3
1	0.000	1.00000		-0.5 0.731531	0.0	0.25	1.00
2	0.250	1.26840		- 3 ,	8.	3a. 2''	0.0
3	1.000	1.718282	0.0	00	0,	0.8	

• Soln. There are 4 data points. Three intervals are there *i* = 1,2,3

$$\begin{split} f_i(x) &= a_i + b_i x + c_i x^2 + d_i x^3 \\ &= \frac{(x_i - x)^3}{6(x_i - x_{i-1})} f_{i-1} " + \frac{(x - x_{i-1})^3}{6(x_i - x_{i-1})} f_i " + (x_i - x) \left\{ \frac{1}{(x_i - x_{i-1})} f_{i-1} - \frac{(x_i - x_{i-1})}{6} f_{i-1} " \right\} \\ &+ (x - x_{i-1}) \left\{ \frac{f_i}{(x_i - x_{i-1})} - \frac{(x_i - x_{i-1})}{6} f_i " \right\} \end{split}$$

We have
$$f_0 = f_3 = 0.0$$

 \therefore For $i = 1$ interval, we have $0.5 \le x \le 0$
i.e. $(x_1 - x_0)f_0 + 2(x_2 - x_0)f_1 + (x_2 - x_1)f_2 = 6\frac{f_2 - f_1}{x_2 - x_1} - 6\frac{f_1 - f_0}{x_1 - x_0}$
i.e. $0.5 \times 0.0 + 2 \times 0.95 \times f_1 + 0.25 \times f_2 = 6 \times \frac{1.2684 - 1}{0.25} - 6 \times \frac{1 - 0.731531}{0.5}$
i.e. $1.50f_1 + 0.25f_2 = 3.219972$
Similarly, for interval $i = 2, \ 0 \le x \le 0.25$
We have,
 $(x_2 - x_1)f_1 + 2(x_3 - x_1)f_2 + (x_3 - x_2)f_3 = 6\frac{f_3 - f_2}{2} - 6\frac{f_2 - f_1}{2}$

 $x_3 - x_2$ $x_2 - x_1$ i.e. $0.25 \times f_1$ "+ 2×1× f_2 "+ 0.75×0.0 = -2.842544.



Similarly obtain cubic polynomial for interval i = 2 and 3.