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1. INTRODUCTION

The spline-under-tension, developed by Schweikert¹⁴ and Cline², introduces a parameter which gives some control on the shape of the spline curve. The tension spline involves the use of hyperbolic functions and can be considered within a general setting proposed by Pruess¹³, where other alternatives are discussed. In particular a rational spline due to Späth¹⁵ is considered.

For parametric representations, Nielson¹¹ describes a polynomial alternative to the spline-under-tension. Here use is made of the additional freedom given by relaxing C² parametric continuity to that of 'geometric' or 'visual' C² continuity. This idea has more recently been taken up by Barsky and Beatty¹.

The specific problem of shape preserving interpolation has been considered by a number of authors. Fritsch and Carlson⁶ and Fritsch and Butland⁵ discuss the interpolation of monotonic data using C¹ piecewise cubic polynomials. McAllister, Passow and Roulier⁹ and Passow and Roulier¹² consider the problem of interpolating monotonic and convex data. They make use of piecewise polynomial Bernstein-Bezier representations and introduce additional knots into their schemes. In particular McAllister and Roulier¹⁰ describe an algorithm for quadratic spline interpolation.

In this paper we will discuss a rational spline solution to the problem of shape preserving interpolation based on references 3, 4, 7 and 8. The rational spline is represented in terms of first derivative values at the knots and provides an alternative to the spline-under-tension. The idea of making the shape control parameters dependent on the first derivative unknowns is then explored. We are then able to preserve the monotonic or convex shape of the interpolation data automatically through the solution of the resulting non-linear consistency equations of the spline.

2. A RATIONAL ALTERNATIVE TO THE SPLINE-UNDER-TENSION

Let (x_i, f_i) , i = 1, ..., n, be a given set of real data, where $x_1 < x_2 < ... < x_n$ and let d_i , i = 1, ..., n, denote first derivative values defined at the knots x_i , i = 1, ..., n. A function $s \in C^1[x, x_n]$ such that

(2.1)
$$s(x_i) = f_i$$
 and $s^{(1)}(x_i) = d_i$, $i = 1, ..., n$

is piecewise defined for $x \in [x_i, x_{i+1}]$, i = 1, ..., n-1 by $s(x) = s_i(x; r_i)$, where

(2.2)
$$s_i(x;r_i) = \frac{(1-\theta)^2(1-\theta+r_i\theta)f_i+(1-\theta)^2\theta h_i d_i^2-\theta^2(1-\theta)h_i d_{i+1}^2+\theta^2(\theta+r_i^2(1-\theta))f_{i+1}}{1+(r_i^2-3)\theta(1-\theta)}$$

and

(2.3)
$$\theta = (x - x_i) / h_i$$
, $h_i = x_{i+1} - x_i$, $r_i > -1$.

The parameters r_i will be used to control the shape of the curve s. The case $r_i = 3$ is that of piecewise cubic Hermite interpolation.

The C^2 spline constraints

(2.4)
$$s^{(2)}(x_{i-}) = s^{(2)}(x_{i+})$$
, $i = 2, ..., n-1$,

give the consistency equations

(2.5)
$$h_i d_{i-1} + [h_i (r_{i-1} - 1) + h_{i-1} (r_i - 1)] d_i + h_{i-1} d_{i+1}$$

= $h_i r_{i-1} \Delta_{i-1} + h_{i-1} r_i \Delta_i$, $i = 2, ..., n-1$,

where

(2.6)
$$\Delta_i = (f_{i+1} - f_i)/h_i$$
, $i = 1, \dots, n-1$

and we assume that d_1 and d_n are given as end conditions. The case $r_i = 3$ corresponds to that of cubic spline interpolation.

Assume that

(2.7)
$$r_i \ge r > 2$$
, $i = 1, ..., n-1$.

Then the linear system (2.5) is strictly diagonally dominant and hence has a unique solution. The solution is also bounded with respect to the r_i since

(2.8)
$$\max |d_i| \leq [r/(r-2)] \max |\Delta_i|.$$

The rational cubic (2.2) can be written as

(2.9)
$$s_i(x;r_i) = \ell_i(x) + e_i(x;r_i)$$

where l_i is the linear interpolant

(2.10)
$$\ell_{i}(x) = (1-\theta)f_{i} + \theta f_{i+1}$$

and

(2.11)
$$e_{i}(x;r_{i}) = \frac{\theta(1-\theta)[(2\theta-1)(f_{i+1}-f_{i}) + (1-\theta)h_{i}d_{i} - \theta h_{i}d_{i+1}]}{1 + (r_{i}-3)\theta(1-\theta)}$$

Thus, as the parameters r_i are increased, it can be shown that the rational spline s converges uniformly to a piecewise defined linear interpolant. An identical argument applies to the rational spline representation of parametric curves.

Suppose $f_i = f(x_i)$ and $f_{i+1} = f(x_{i+1})$, where $f \in C^4[x_i x_{i+1}]$. Then an error bound for the rational cubic on $[x_i, x_{i+1}]$ is given by

$$(2.12) |f(x) - s_{i}(x;r_{i})| \leq \frac{h_{i}}{4c_{i}} \max \left\{ |f_{i}^{(1)} - d_{i}|, |f_{i+1}^{(1)} - d_{i+1}| \right\} \\ + \frac{1}{384c_{i}} \left[h_{i}^{4} || f^{(4)} || + |r_{i} - 3| \left\{ \frac{1}{4} h_{i}^{4} || f^{(4)} || + 4h_{i}^{3} || f^{(3)} || + 12h_{i}^{2} || f^{(2)} || \right\} \right],$$

where

$$c_i = \min\{1, (1+r_i)/4\}$$
 and $||f|| = \max_{x \in [x_i, x_{i+1}]} |f(x)|$.

This result will influence the choice of the parameters r_i when the interpolation data are monotonic or convex. In particular we wish to choose r_i such that $r_i - 3 = O(h_i^2)$, whilst maintaining monotonicity or convexity of the interpolant.

3. THE INTERPOLATION OF MONOTONIC DATA

Suppose that

(3.1) $f_1 < f_2 < \ldots < f_n$ (monotonic increasing data)

and assume the derivative values ${\tt d}_{\underline{i}}$ satisfy the necessary monotonicity conditions

$$(3.2) d_i > 0, i = 2,...,n-1,$$

where $d_1 > 0$ and $d_n > 0$ are given. Then

(3.3)
$$r_i = 1 + (d_i + d_{i+1}) / \Delta_i$$

ensures that $s_i^{(1)}(x;r_i) > 0$ on $[x_i,x_{i+1}]$ and hence the rational cubic is monotonic increasing.

Substituting (3.3) into the C^2 consistency equations (2.5) gives the non-linear system

$$(3.4) \quad d_{i}[-c_{i} + a_{i-1}d_{i-1} + (a_{i-1} + a_{i})d_{i} + a_{i}d_{i+1}] = b_{i}, \quad i = 2, \dots, n-1,$$

where

(3.5)
$$a_i = 1/(h_i \Delta_i)$$
, $b_i = \Delta_{i-1}/h_{i-1} + \Delta_i/h_i$, $c_i = 1/h_{i-1} + 1/h_i$.

This system has a unique solution satisfying the monotonicity conditions (3.2). The "Gauss-Seidel" iteration

(3.6)
$$d_{i}^{(k+1)} = G_{i}(d_{i-1}^{(k+1)}, d_{i+1}^{(k)})^{*}, \quad i = 2, ..., n-1$$

where

(3.7)
$$G_{i}(\xi,\eta) = \frac{1}{2(a_{i-1}+a_{i})} \{ \zeta + [\zeta^{2} + 4(a_{i-1}+a_{i})b_{i}]^{\frac{1}{2}} \},$$

(3.8)
$$\zeta = c_i - a_{i-1}\xi - a_i\eta$$
,

provides a robust algorithm for solving (3.4), being globally convergent to the required positive solution. The method also provides an $O(h^4)$ accurate solution with reference to the error bound (2.12).

4. THE INTERPOLATION OF CONVEX DATA

Let

$$(4.1) \qquad \qquad \Delta_1 < \Delta_2 < \ldots < \Delta_{n-1} \quad (\text{convex data}) ,$$

and assume the derivative values satisfy the necessary convexity conditions

(4.2)
$$\Delta_{i-1} < d_i < \Delta_i$$
, $i = 2, ..., n-1$,

where $d_1 < \Delta_1$ and $d_n > \Delta_{n-1}$ are given. Then

(4.3)
$$r_i = 1 + (d_{i+1} - \Delta_i) / (\Delta_i - d_i) + (\Delta_i - d_i) / (d_{i+1} - \Delta_i)$$

ensures that $s_i^{(2)}(x;r_i) > 0$ on $[x_i,x_{i+1}]$ and hence the rational cubic is convex.

The C² consistency equations are now

(4.4)
$$\left(\frac{\Delta_{i}-d_{i}}{d_{i}-\Delta_{i-1}}\right)^{2} = \frac{h_{i}}{h_{i-1}}\left(\frac{d_{i+1}-\Delta_{i}}{\Delta_{i-1}-d_{i-1}}\right), \quad i = 2, \dots, n-1$$

giving a non-linear system with a unique solution satisfying the convexity conditions (4.2). This solution can be found using a "Gauss-Seidel" iteration as in (3.6), where we now take

(4.5)
$$G_{i}(\xi,\eta) = \frac{h_{i-1}^{\frac{1}{2}}(\Delta_{i-1} - \xi)^{\frac{1}{2}}\Delta_{i} + h_{i}^{\frac{1}{2}}(\eta - \Delta_{i})^{\frac{1}{2}}\Delta_{i-1}}{h_{i-1}^{\frac{1}{2}}(\Delta_{i-1} - \xi)^{\frac{1}{2}} + h_{i}^{\frac{1}{2}}(\eta - \Delta_{i})^{\frac{1}{2}}}$$

and choose intial values $d_i^{(0)}$, i = 2,...,n-1, for the iteration such that the residuals alternate. Finally, the convex spline method like the monotonic spline is $O(h^4)$ accurate.

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