Weighted Nu Splines

Abstract. Weighted ν-splines are the composition of two spline methods, namely, weighted splines and ν-splines. These are the generalization of cubic spline method and are highly useful for CAD/CAM and various applications in computer graphics. Both—interpolatory and freeform—schemes are available in the literature. This chapter explains interpolatory weighted ν-splines together with a construction of its B-spline-like form. The design curves, constructed through B-spline-like form, possess all the ideal geometric properties such as partition of unity, convex hull, and variation diminishing. The splines provide not only a variety of very interesting shape control such as point and interval tensions but also, as a special case, recover the cubic spline method. In addition, these weighted ν-splines also provide, as special cases, the weighted splines and the ν-splines. The method for evaluating these splines is suggested by a transformation to Bézier form.

2.1 Introduction

Designing of curves, especially those curves that are robust and easy to control and compute, has been one of the significant problems of computer graphics and geometric modeling. Specific applications including font designing, capturing hand-drawn images on computer screens, data visualization, and computer-supported cartooning are main motivations toward curve designing. In addition, various other applications in CAD/CAM/CAGD are also a good reason to study this topic. Many authors have worked in this direction. For brevity, the reader is referred to [1–22].

A cubic spline curve method is considered to be a considerably decent approach for designing applications in the area of computer graphics and geometric modeling. However, due to its various limitations, such as lack of freedom in shape control, a designer may not have much help. In this study, the weighted ν-spline method has been reviewed. This curve design method, in addition to enjoying the good features of cubic splines, possesses interesting shape design features too. It has two families of shape parameters working in such a way that one family of parameters is associated with intervals and the other with points. These parameters provide a variety of shape control such as point and interval tension. This is an interpolatory curve scheme, which utilizes a piecewise cubic function in its description. However, it is desired to extend this idea to freeform curves, which
can enjoy all the ideal properties related to B-spline theory. This work is mainly concerned with developing such a theory.

Weighted splines [7] were discovered as a cubic spline method. The method provides a $C^1$ computationally simpler alternative to the exponential spline-under-tension [4, 13, 20]. Regarding shape characteristics, it has shape control parameters associated with each interval, which can be used to flatten or tighten the curve locally. Nu-splines [11, 12] were discovered as another cubic spline method. It provides a $G^C_2$ computationally simpler alternative to the exponential spline-under-tension [4, 13, 21]. Regarding shape characteristics, it has shape control parameters associated with each point, which can be used to tighten the curve both locally and globally. The ideas of weighted splines and Nu-splines were married together to formulate another spline called weighted Nu-spline [11, 12, 19, 22]. This curve design method covers the shape features of both of its counter parts and provides a $C^1$ computationally economical method.

B-splines are a useful and powerful tool for computer graphics and geometric modeling. They can be found frequently in the existing CAD/CAM (computer-aided design/computer-aided manufacturing) systems. They form a basis for the space of $n$ th degree splines of continuity class $C^{n-1}$. Each B-spline is a non-negative $n$ th-degree spline that is nonzero only on $n + 1$ intervals. The B-splines form a partition of unity, that is, they sum up to one. Curves generated by summing control points multiplied by the B-splines have some very desirable shape properties, including the local convex hull property and variation diminishing property.

It is desirable to generalize the idea of B-spline-like local basis functions for the classes of splines with shape parameters considered in the description of continuity. The first local basis for $G^C_2$ splines was developed by Lewis [10]. In 1981, Barsky [1] generalized B-splines to Beta splines. These splines preserve the geometric smoothness of the design curve while allowing the continuity conditions on the spline functions at the knots to be varied by certain parameters, thus giving greater flexibility. Later, in 1984, Bartels and Beatty [2] developed local bases for Beta spline curves that are equivalent to Boehm’s [3] Gamma splines. Foley [7], in 1987, constructed a B-spline-like basis for weighted splines; different weights were built into the basis functions so that the control point curve was a $C^1$ piece-wise cubic with local control of interval tension.

In this work, a constructive approach has been adopted to build B-spline-like basis for cubic spline curves with the same continuity constraints as those for interpolatory weighted $v$-splines. These are local basis functions with local support which have the property of being positive everywhere. The design curve, constructed through these functions, possesses all the ideal geometric properties like partition of unity, convex hull, and variation diminishing. This curve method provides not only a variety of very interesting shape control such as point and interval tensions, but also, as a special case, recovers the cubic B-spline curve method. In addition, it also provides B-spline-like design curves for weighted splines, $v$-splines and weighted $v$-splines. The method for evaluating these splines is suggested by a transformation to Bézier form.
The approach adopted in the construction of local basis for the weighted $\nu$-splines is quite different from those adopted for different spline methods in [1–8, 15]. The way for evaluating the weighted $\nu$-splines representation of a curve is suggested by a transformation to piecewise defined Bézier form. This form will also expedite a proof of the variation diminishing property for the Bézier representation.

This chapter is related to the weighted $\nu$-spline method [19, 22] explained in Section 2.2. It studies, in Section 2.3, a B-spline-like local basis for the weighted $\nu$-spline. The design curve in Section 2.4 maintains the $C^1$ continuity of the weighted $\nu$-splines. This description of freeform weighted $\nu$-spline not only provides a variety of interesting shape control such as point and interval tensions but also, as a special case, recovers the cubic B-spline curve method. In addition, it also provides B-spline like design curves for weighted splines, $\nu$-splines and weighted $\nu$-splines. The method has been extended for the construction of surfaces in Section 2.5. Section 2.6 summarizes the chapter.

2.2 Some Spline Methods

This section gives a brief review of the cubic spline, weighted splines, $\nu$-splines, and weighted $\nu$-splines. Detailed description of the weighted $\nu$-splines is given in Sections 2.3 and 2.4. Assume that we are given knot partition as $t_1 < t_2 < \ldots < t_n$, and set of control points $F_1, F_2, \ldots, F_n$. Let us have the followings:

Point tension factors: $v_i \geq 0, i = 1, 2, \ldots, n$,

Interval weights: $w_i > 0, i = 1, 2, \ldots, n$, \(i=1,2,\ldots,n\). \(2.1\)

Consider the piecewise cubic function:

$$p(t) = p_i(t) = F_i(1 - \theta)^3 + 3\theta(1 - \theta)^2V_i + 3\theta^2(1 - \theta)W_i + F_{i+1}\theta^3, \quad 2.2$$

where

$$\theta = \frac{t - t_i}{h_i}, \quad h_i = t_{i+1} - t_i, \quad 2.3$$

and

$$V_i = F_i + \frac{h_iD_i}{3}, \quad W_i = F_{i+1} - \frac{h_iD_{i+1}}{3}. \quad 2.4$$

It is obvious to see that the piecewise cubic function (2.2) holds the following interpolatory properties:

$$p(t_i) = F_i, \quad p(t_{i+1}) = F_{i+1}, \quad 2.5$$

$$p^{(1)}(t_i) = D_i, \quad p^{(1)}(t_{i+1}) = D_{i+1}$$

where $p^{(1)}$ denotes first derivative with respect to $t$ and $D_i$ denote derivative values given at the knots $t_i$. This leads the piecewise cubic (2.2) to the piecewise Hermite interpolant $p \in C^1[t_1, t_n]$. 

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2.2.1 Cubic Splines

The cubic spline interpolant is a \( C^2 \) piecewise cubic function \( p(t) \) that minimizes

\[
V(f) = \sum_{i=1}^{n-1} \int_{t_i}^{t_{i+1}} [f''(t)]^2 dt,
\]

subject to the interpolation conditions \( f(t_i) = F_i \) for \( i = 1, 2, \ldots, n \) and one of the following end conditions:

- **Type 1:** First derivative end conditions,
- **Type 2:** Natural end conditions, or
- **Type 3:** Periodic end conditions.

Given \( F_i \) and \( D_i \) for \( i = 1, 2, \ldots, n \), there exists a unique \( C^2 \) piecewise cubic function \( f(t) \) that satisfies

\[
f(t_i) = F_i \quad \text{and} \quad f'(t_i) = D_i \quad \text{for} \quad i = 1, 2, \ldots, n.
\]

The unknowns are the first derivative values, \( D_i \), \( i = 1, 2, \ldots, n \), and once they are computed, the function \( f(t) \) can be easily evaluated using the standard piecewise cubic Hermite form explained in (2.2). Necessary and sufficient conditions for the function \( f(t) \) to be the cubic spline interpolant are that its derivatives \( D_i \)’s satisfy

\[
\hat{c}_{i-1} D_{i-1} + (2 \hat{c}_i + 2 \hat{c}_i) D_i + \hat{c}_i D_{i+1} = \hat{b}_i (F_{i+1} - F_i) + \hat{b}_{i-1} (F_i - F_{i-1}),
\]

for \( i = 1, 2, \ldots, n \), where \( \hat{c}_i = 1/h_i, \hat{b}_i = 3 \hat{c}_i/h_i \). The above system of equations provides \( (2n-2) \) equations for \( n \) unknowns, \( D_1, \ldots, D_n \), and the additional equations come from the given end conditions. The equations for Type I first derivative end conditions are

\[
2 \hat{c}_1 D_1 + \hat{c}_1 D_2 = \hat{b}_1 (F_2 - F_1),
\]

and

\[
\hat{c}_{n-1} D_{n-1} + 2 \hat{c}_n D_n = \hat{b}_{n-1} (F_n - F_{n-1}).
\]

For Type 3 periodic end conditions, they are

\[
(2 \hat{c}_1 + 2 \hat{c}_n - 1) D_1 + \hat{c}_1 D_2 + \hat{c}_{n-1} D_{n-1} = \hat{b}_1 (F_2 - F_1) + \hat{b}_{n-1} (F_n - F_{n-1}),
\]

and \( D_1 = D_n \). The linear system of equations that occurs when Type 1 or 2 end conditions are used is tridiagonal and diagonally dominant; thus it can be solved efficiently by using a standard tridiagonal system solver. Figure 2.1 is a cubic spline curve for a data shown as bullets.

2.2.2 Weighted Splines

The weighted spline interpolant is a \( C^1 \) piecewise cubic function \( p(t) \) that minimizes

\[
V(f) = \sum_{i=1}^{n-1} \int_{t_i}^{t_{i+1}} [f''(t)]^2 dt,
\]
subject to the interpolation conditions \( f(t_i) = F_i \) for \( i = 1, 2, \ldots, n \) and one of the Type 1, Type 2, and Type 3 end conditions.

The \( \omega_i \)'s are termed as interval weights because they “tighten” the curve on the \( i \)th interval in the same way that they do for the weighted splines in [14]. If and all \( \omega_i = q \), where \( q \) is some constant value, then the weighted spline equals the cubic spline as in Section 2.2.1.

The approach taken in [21] uses piecewise cubic Hermite basis functions to represent the weighted splines. Given \( F_i \) and \( D_i \) for \( i = 1, 2, \ldots, n \), there exists a unique \( C^1 \) piecewise cubic function \( f(t) \) that satisfies \( f(t_i) = F_i \) and \( f'(t_i) = D_i \) for \( i = 1, 2, \ldots, n \). The unknowns are the first derivative values, \( D_i, i = 1, 2, \ldots, n \), and once they are computed, the function \( f(t) \) can be easily evaluated using the standard piecewise cubic Hermite form. Necessary and sufficient conditions for the function \( p(t) \) to be the weighted spline interpolant are that its derivatives \( D_i \) satisfy

\[
c_{i-1}D_i - (2c_{i-1} + 2c_i)D_i + c_iD_{i+1} = b_i (F_{i+1} - F_i) + b_{i-1} (F_i - F_{i-1}),
\]

for \( i = 1, 2, \ldots, n \), where \( c_i = \omega_i / h_i \), \( b_i = 3c_i / h_i \). The above system of equations provides \( (n - 2) \) equations for \( n \) unknowns, \( D_1, \ldots, D_n \), and the additional equations come from the given end conditions. The equations for Type I first derivative end conditions are \( D_1 = f'(t_1) \) and \( D_n = f'(t_n) \). For Type II natural end conditions they are

\[
2c_1D_1 + c_1D_2 = b_1 (F_2 - F_1),
\]

and

\[
c_n D_{n-1} + 2c_{n-1}D_n = b_{n-1}(F_n - F_{n-1}).
\]

For Type 3 periodic end conditions, they are

\[
2c_1 + 2c_{n-1}D_1 + c_1D_2 + c_{n-1}D_{n-1} = b_1(F_2 - F_1) + b_{n-1}(F_n - F_{n-1}),
\]
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and $D_1 = D_n$. The linear system of equations that occurs when Type 1 or 2 end conditions are used is tridiagonal and diagonally dominant; thus it can be solved efficiently by using a standard tridiagonal system solver.

2.2.3 Nu Splines

The $\nu$-spline interpolant is a $GC^2$ piecewise cubic function $p(t)$ that minimizes

$$V(f) = \sum_{i=1}^{n} v_i \left[f'(t_i)\right]^2,$$

subject to the interpolation conditions $f(t_i) = F_i$ for $i = 1, 2, \ldots, n$ and one of the Type 1, Type 2, and Type 3 end conditions.

The $v_i$ are termed point tension factors because they “tighten” a parametric curve at the $i$th point in the same way that they do for the $\nu$-splines in [11, 12]. If $v_i = 0$ $\nu$-spline equals the cubic spline in [11, 12].

The approach taken in [11, 12, 19, 21] uses piecewise cubic Hermite basis functions to represent the $\nu$-splines. Given $F_i$ and $D_i$ for $i = 1, 2, \ldots, n$, there exists a unique $GC^2$ piecewise cubic function $f(t)$ that satisfies $f(t_i) = F_i$ and $f'(t_i) = D_i$ for $i = 1, 2, \ldots, n$. The unknowns are the first derivative values, $D_i$, $i = 1, 2, \ldots, n$, and once they are computed, the function $f(t)$ can be easily evaluated using the standard piecewise cubic Hermite form. Necessary and sufficient conditions for the function $p(t)$ to be the $\nu$-spline interpolant are that its derivatives $D_i$ satisfy

$$\ddot{c}_{i-1}D_{i-1} + \left(\frac{1}{2}v_i + 2\dot{c}_{i-1} + 2\ddot{c}_i\right)D_i + \ddot{c}_iD_{i+1} = \ddot{b}_i (F_{i+1} - F_i) + \ddot{b}_{i-1} (F_i - F_{i-1}),$$

for $i = 1, 2, \ldots, n$, where $\ddot{c}_i = 1/h_i$, $\dddot{b}_i = 3\dddot{c}_i/h_i$. The above system of equations provides $(n - 2)$ equations for $n$ unknowns, $D_1, \ldots, D_n$, and the additional equations come from the given end conditions. The equations for Type I first derivative end conditions are $D_1 = f'(t_1)$ and $D_n = f'(t_n)$. For Type II natural end conditions they are

$$\left(\frac{1}{2}v_1 + 2c_1\right)D_1 + c_1 D_2 = b_1 (F_2 - F_1),$$

and

$$c_{n-1}D_{n-1} + \left(\frac{1}{2}v_n + 2c_{n-1}\right)D_n = b_{n-1}(F_n - F_{n-1}).$$

For Type 3 periodic end conditions, they are

$$\left(\frac{1}{2}v_1 + \frac{1}{2}v_n + 2c_1 + 2c_{n-1}\right)D_1 + c_1 D_2 + c_{n-1} D_{n-1}$$

$$= b_1(F_2 - F_1) + b_{n-1}(F_n - F_{n-1}),$$

and $D_1 = D_n$. The linear system of equations that occurs when Type 1 or 2 end conditions are used is tridiagonal and diagonally dominant; thus it can be solved efficiently by using a standard tridiagonal system solver.
2.2.4 Weighted Nu Splines

The weighted ν-spline interpolant is a $C^1$ piecewise cubic function $p(t)$ that minimizes

$$V(f) = \sum_{i=1}^{n} \omega_i \int_{t_i}^{t_{i+1}} [f''(t)]^2 dt + \sum_{i=1}^{n} \nu_i [f'(t_i)]^2,$$

subject to the interpolation conditions $f(t_i) = F_i$ for $i = 1, 2, \ldots, n$ and one of the Type 1, Type 2, Type 3 end conditions. This is the marriage of Weighted splines and Nu splines which can be recovered as special cases discussed later in this chapter.

The $\nu_i$ are termed point tension factors because they ‘tighten’ a parametric curve at the $i$th point in the same way that they do for the ν-splines in [11, 12]. The $\omega_i$ are termed interval weights because they ‘tighten’ the curve on the $i$th interval in the same way that they do for the weighted splines in [14]. If $\nu_i = 0$ and all $\omega_i = q$, where $q$ is some constant value, then the weighted ν-spline equals the ν-spline in [11, 12] with tension factors $\nu_i/q$. If all $\nu_i = 0$, then it equals the weighted spline given in [14].

The approach taken in [8] uses piecewise cubic Hermite basis functions to represent the weighted ν-splines. Given $F_i$ and $D_i$ for $i = 1, 2, \ldots, n$, there exists a unique $C^1$ piecewise cubic function $f(t)$ that satisfies $f(t_i) = F_i$ and $f'(t_i) = D_i$ for $i = 1, 2, \ldots, n$. The unknowns are the first derivative values, $D_i$, $i = 1, 2, \ldots, n$, and once they are computed, the function $f(t)$ can be easily evaluated using the standard piecewise cubic Hermite form. Necessary and sufficient conditions for the function $p(t)$ to be the weighted ν-spline interpolant are that its derivatives $D_i$ satisfy

$$c_{i-1} D_{i-1} + \left( \frac{1}{2} \nu_i + 2c_{i-1} + 2c_i \right) D_i + c_i D_{i+1} = b_i (F_{i+1} - F_i) + b_{i-1} (F_i - F_{i-1}),$$

for $i = 1, 2, \ldots, n$. The above system of equations provides $(n-2)$ equations for $n$ unknowns, $D_1, \ldots, D_n$, and the additional equations come from the given end conditions. The equations for Type I first derivative end conditions are $D_1 = f'(t_1)$ and $D_n = f'(t_n)$. For Type II natural end conditions they are

$$\left( \frac{1}{2} \nu_1 + 2c_1 \right) D_1 + c_1 D_2 = b_1 (F_2 - F_1),$$

and

$$c_{n-1} D_{n-1} + \left( \frac{1}{2} \nu_n + 2c_{n-1} \right) D_n = b_{n-1} (F_n - F_{n-1}).$$

For Type 3 periodic end conditions, they are

$$\left( \frac{1}{2} \nu_1 + \frac{1}{2} \nu_n + 2c_1 + 2c_{n-1} \right) D_1 + c_1 D_2 + c_{n-1} D_{n-1} = b_1 (F_2 - F_1) + b_{n-1} (F_n - F_{n-1}),$$
and $D_1 = D_n$. The linear system of equations that occurs when Type 1 or 2 end conditions are used is tridiagonal and diagonally dominant; thus it can be solved efficiently by using a standard tridiagonal system solver.

The weighted $\nu$-spline can be computed by solving for $D_i$’s. This can be done by re-writing the system of equations in (2.10) as follows:

$$c_{i-1}D_{i-1} + \left( \frac{\nu_i}{2} + 2c_{i-1} + 2c_i \right) D_i + c_i D_{i+1} = 3c_i \Delta_i + 3c_{i-1} \Delta_{i-1}, \quad (2.7)$$

where

$$\Delta_i = (F_{i+1} - F_i) / h_i,$$

for $i = 2, \ldots, n-1$. For given appropriate end conditions (Type 1, Type 2, or Type 3), this system of equations is a tridiagonal linear system. This is also diagonally dominant for the following constraints on the shape parameters as in (2.1), and hence has a unique solution for $D_i$’s. As far as the computation method is concerned, it is much more economical to adopt the LU-decomposition method to solve the tridiagonal system. Therefore, the above discussion can be concluded in the following:

**Theorem 2.1.** For the shape parameter constraints (2.1), the spline solution of the weighted $\nu$-spline exists and is unique.

**Remark 2.1.** Each component of the parametric weighted $\nu$-spline is a $C^1$ function in general, but it has second-order geometric continuity at $t_i$ if $\omega_{i-1} = \omega_i$ and the tangent vector at $t_i$ is non-zero and it is $C^2$ at $t_i$ if $\omega_{i-1} = \omega_i$ and $\nu_i = 0$.

### 2.2.5 Demonstration

Figure 2.1 is the parametric weighted $\nu$-spline interpolant to the points denoted by circles using periodic end conditions. In Figure 2.2, interval weight, $\omega_i$, of 30 is used in the base interval, while point tension factors, $\nu_i$ of 10 are used on the four vertices defining the “neck.” The rest of the parameters are taken as $\omega_i = 1$ and $\nu_i = 0$.

### 2.3 Freeform Weighted Nu Spline

This section is devoted to constructing the freeform weighted Nu spline which has inherent properties of B-spline curves. This formulation is possible through the construction of local support basis $B_i$’s to compute the cubic weighted $\nu$-spline $p(t)$ satisfying the following constraints:

$$\begin{bmatrix} p(t_{i+}) \\ p^{(1)}(t_{i+}) \\ p^{(2)}(t_{i+}) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & \frac{\nu_i}{\omega_i} & \frac{\omega_{i-1}}{\omega_i} \end{bmatrix} \begin{bmatrix} p(t_{i-}) \\ p^{(1)}(t_{i-}) \\ p^{(2)}(t_{i-}) \end{bmatrix}, \quad (2.8)$$
2.3. Freeform Weighted Nu Spline

2.3.1 Local Support Basis

For the purpose of the analysis, let additional knots be introduced outside the knot partition \( t_1 < t_2 < \ldots < t_n \) of the interval \([t_1, t_n]\), defined by:

\[
 t_{-2} < t_{-1} < t_0 < t_1 \quad \text{and} \quad t_n < t_{n+1} < t_{n+2} < t_{n+3}.
\] (2.9)

Let

\[
a_i = \frac{1}{c_i},
\] (2.10)

and \( \phi_i \) be cubic weighted \( \nu \)-spline:

\[
\phi_i (t) = \begin{cases} 
0, & t \leq t_{i-2}, \\
1, & t \geq t_{i+1}.
\end{cases}
\] (2.11)

Imposing weighted \( \nu \)-spline constraints (2.8), we have:

\[
\phi_i (t_{i-1}) = \frac{h_{i-2}}{3} \phi_i^{(1)} (t_{i-1}),
\]

\[
\phi_i (t_i) = 1 - \frac{h_i}{3} \phi_i^{(1)} (t_i),
\]

\[
\phi_i^{(1)} (t_{i-1}) = \frac{A_i}{C_i},
\]

and

\[
\phi_i^{(1)} (t_i) = \frac{B_i}{C_i},
\]
where, if \( d_i = \frac{1}{2} a_i a_{i-1} v_i + a_{i-1} + a_i \), then

\[
A_i = \frac{3a_{i-2}}{h_{i-1}} d_i,
\]

\[
B_i = \frac{3a_i}{h_{i-1}} d_{i-1},
\]

\[
C_i = d_i d_{i-1} + \frac{a_i}{h_{i-1}} (h_{i-1} + h_i) d_{i-1} + \frac{a_{i-2}}{h_{i-1}} (h_{i-1} + h_{i-2}) d_i.
\]

Let

\[
D_i = h_{i-1} d_i d_{i-1} + a_i (h_{i-1} + h_i) d_{i-1} + a_{i-2} (h_{i-1} + h_{i-2}) d_i,
\]

\[
\mu_i = \phi_{i+1}(t_i), \quad \lambda_i = 1 - \phi_i(t_i),
\]

\[
\hat{\mu}_i = \phi_{i+1}^{(1)}(t_i), \quad \hat{\lambda}_i = \phi_i^{(1)}(t_i),
\]

Then

\[
\hat{\lambda}_i = \frac{3 a_i d_{i-1}}{D_i}, \quad \hat{\mu}_i = \frac{3 a_{i-1} d_{i+1}}{D_{i+1}},
\]

\[
\mu_i = \frac{h_{i-1}}{3} \hat{\mu}_i, \quad \lambda_i = \frac{h_i}{3} \hat{\lambda}_i,
\]

and hence

\[
0 \leq \mu_i \leq 1, \quad 0 \leq \lambda_i \leq 1 \quad \text{and} \quad 0 \leq \mu_i + \lambda_i \leq 1.
\]

Now define

\[
B_i(t) = \phi_i(t) - \phi_{i+1}(t).
\]

Then \( B_i \) has the local support \((t_{i-2}, t_{i+2})\) and an explicit representation of \( B_j \) on any interval \((t_i, t_{i+1})\) (in particular, for \( i = j - 2, j - 1, j, j + 1 \)) can be calculated as:

\[
B_j(t) = (1 - \theta)^3 B_j(t_i) + \theta (1 - \theta)^2 (3 B_j(t_i) h_i B_j^{(1)}(t_i)) + \theta^2 (1 - \theta) (3 B_j(t_{i+1}) - h_i B_j^{(1)}(t_{i+1})) + \theta^3 B_j(t_{i+1}), \quad (2.12)
\]

where

\[
B_j(t_i) = B_j^{(1)}(t_i) = 0 \quad \text{for} \quad i \neq j - 1, j, j + 1,
\]

and

\[
\begin{align*}
B_j(t_{j-1}) &= \mu_{j-1}, & B_j^{(1)}(t_{j-1}) &= \hat{\mu}_{j-1}, \\
B_j(t_j) &= 1 - \lambda_j - \mu_j, & B_j^{(1)}(t_j) &= \hat{\lambda}_j - \hat{\mu}_j, \\
B_j(t_{j+1}) &= \lambda_{j+1}, & B_j^{(1)}(t_{j+1}) &= -\hat{\lambda}_{j+1}.
\end{align*}
\]

(2.13)

Careful examination of the Bézier vertices of \( B_j(t) \) in (2.12) shows these to be non-negative for \( v_i, \omega_i \) satisfying (2.7) and thus \( B_j(t) \geq 0 \), \( \forall t \). This leads to the following:
Proposition 2.1. The local support basis functions (2.12) are such that the following properties hold:

(i) (Local support) \( B_j(t) = 0 \) for \( t \notin (t_{j-2}, t_{j+2}) \).

(ii) (Partition of unity) \( \sum_{j=-1}^{n+1} B_j(t) = 1 \) for \( t \in [t_1, t_n] \).

(iii) (Positivity) \( B_j(t) \geq 0 \) for all \( t \).

2.3.2 Design Curve

Now, we need a convenient method to compute the curve representation. It is desired to apply the above developed local basis functions to develop a freeform weighted \( \nu \)-spline curve as follows:

\[
P(t) = \sum_{j=-1}^{n+1} B_j(t) P_j, \quad t \in [t_1, t_n],
\]

where \( P_j \in \mathbb{R}^N, j = 0, 1, \ldots, n+1 \), define the control points of the representation. By the local support property,

\[
P(t) = \sum_{j=-1}^{i+2} B_j(t) P_j, \quad t \in [t_i, t_{i+1}), \quad i = 0, \ldots, n-1.
\]

Substitution of (2.12), \( t \in [t_i, t_{i+1}) \), then gives the piecewise defined Bézier representation

\[
P(t) \equiv P_i(t) = F_i(1 - \theta)^3 + 3\theta(1 - \theta)^2V_i + 3\theta^2(1 - \theta)W_i + F_{i+1}\theta^3, \tag{2.15}
\]

where

\[
\begin{align*}
F_i &= \lambda_i P_{i-1} + (1 - \lambda_i - \mu_i)P_i + \mu_i P_{i+1}, \\
V_i &= (1 - \alpha_i)P_i + \alpha_i P_{i+1}, \\
W_i &= \beta_i P_i + (1 - \beta_i)P_{i+1},
\end{align*}
\]

(2.16)

with

\[
\begin{align*}
\alpha_i &= \mu_i + h_i \hat{\mu}_i / 3 = \frac{\hat{\mu}_i}{3} (h_{i-1} + h_i), \\
\beta_i &= \lambda_{i+1} + h_i \hat{\lambda}_i = \frac{\lambda_{i+1}}{3} (h_i + h_{i+1}).
\end{align*}
\]

This transformation to Bézier form is very convenient for computational purposes and also leads to the following:

Proposition 2.2. (Variation Diminishing Property) The weighted \( \nu \)-spline curve \( P(t), t \in [t_0, t_n] \), defined by (2.14), crosses any (hyper) plane of dimension \( N - 1 \) no more times than it crosses the “control polygon” joining the control points \( P_{-1}, P_0, \ldots, P_n \).
Proof. Following the arguments of positivity in the previous proposition, it is straightforward that $0 \leq \alpha_i \leq 1$, $0 \leq \beta_i \leq 1$, and $0 \leq \alpha_i + \beta_i \leq 1$. Thus, $V_i$ and $W_i$ lie on the line segment joining $P_i$ and $P_{i+1}$, where $V_i$ is before $W_i$. It can also be simply noted that

$$F_i = (1 - \gamma_i)W_{i-1} + \gamma_i V_i,$$  \hspace{1cm} (2.17)

where

$$0 < \gamma_i = \frac{h_{i-1}}{h_{i-1} + h_i} < 1.$$  

Thus, the control polygon of the piecewise defined Bézier representation is obtained by corner cutting of the weighted $\nu$-spline control polygon. Since the piecewise defined Bézier representation is variation diminishing, it follows that weighted $\nu$-spline representation is variation diminishing.

### 2.3.3 Shape Control

The shape parameters, defined in (2.7), can be used to control the local or global shape of the design curve. To analyze such behaviors, the explicit form on $(t_i, t_{i+1})$ of the weighted $\nu$-spline design curve (2.14) can be expressed as:

$$P(t) = l_i(t) + e_i(t),$$  \hspace{1cm} (2.18)

where

$$l_i(t) = (1 - \theta)F_i + \theta F_{i+1},$$  \hspace{1cm} (2.19)

and

$$e_i(t) = \theta(1 - \theta) \left\{ \left[ (F_{i+1} - F_i) - h_i P^{(1)}(t_i) \right] (\theta - 1) + \left[ (F_{i+1} - F_i) - h_i P^{(1)}(t_{i+1}) \right] \theta \right\}. $$  \hspace{1cm} (2.20)

**Proposition 2.3.** Let $\omega_i = \omega \geq 1$, and $v_i = 0, \forall i$ are all bounded then the weighted $\nu$-spline design curve is straightway the standard cubic spline.

**Proof.** It follows from the last constraint of relation (2.8).

**Proposition 2.4.** (Global Tension) Let $\omega_i \geq 1, \forall i$, be bounded and $v_i \geq v$ then the weighted $\nu$-spline curve (2.14) converges uniformly to the control polygon $P_0, \ldots, P_n$ as $\nu \to \infty$.

**Proof.** Let $v_i = v, \forall i$ then from (2.1)

$$\lim_{\nu \to \infty} P^{(1)}(t_i) = 0.$$  \hspace{1cm} (2.21)

Moreover

$$\lim_{\nu \to \infty} \hat{\mu}_i = 0 = \lim_{\nu \to \infty} \hat{\lambda}_i, \forall i.$$
This implies the following:

\[
\lim_{\nu \to \infty} F_i = P_i, \quad \forall i.
\]  
(2.22)

More generally, for \(\nu_i \geq \nu \geq 0\), it can be shown that

\[
\max_i |\hat{\lambda}_i| \leq r(\nu),
\]
and

\[
\max_i |\hat{\mu}_i| \leq s(\nu),
\]

where

\[
\lim_{\nu \to \infty} r(\nu) = 0 = \lim_{\nu \to \infty} s(\nu),
\]

and again (2.21) and (2.22) hold. Hence the result.

**Proposition 2.5.** [Local Tension] Consider an interval \([t_k, t_{k+1}]\) for a fixed \(k\). Then on \([t_k, t_{k+1}]\) weighted \(\nu\)-spline curve converges uniformly to a line segment of the line \(P_k P_{k+1}\) as \(\omega_k \to \infty\) where \(\omega_{k-1}\) and \(\nu_k\) are bounded.

**Proof.** Careful examination shows

\[
\lim_{\omega \to \infty} \mu_k = \frac{h_{k-1}}{(3h_k + h_{k-1} + h_{k+1})} = \hat{\alpha}_k \quad \text{(say)}
\]

\[
\lim_{\omega \to \infty} \mu_{k+1} = 0
\]

\[
\lim_{\omega \to \infty} \lambda_k = 0
\]

\[
\lim_{\omega \to \infty} \lambda_{k+1} = \frac{h_{k+1}}{(3h_k + h_{k-1} + h_{k+1})} = \hat{\beta}_k \quad \text{(say)}
\]

This implies the following:

\[
\lim_{\omega \to \infty} F_k = (1 - \hat{\alpha}_k) P_k + \hat{\alpha}_k P_{k+1} = \hat{F}_k \quad \text{(say)}
\]

and

\[
\lim_{\omega \to \infty} F_{k+1} = \hat{\beta}_k P_k + \left(1 - \hat{\beta}_k\right) P_{k+1} = \hat{F}_{k+1} \quad \text{(say)}
\]

Obviously \(\hat{F}_k\) and \(\hat{F}_{k+1}\) lie on \(T_k P_{k+1}\) and \(\hat{F}_k\) is before \(\hat{F}_{k+1}\) as \(\hat{\alpha}_k < (1 - \hat{\beta}_k)\).

Also

\[
\lim_{\omega \to \infty} (F_{k+1} - F_k) = \lim_{\omega \to \infty} h_k P^{(1)}(t_k) = \lim_{\omega \to \infty} h_k P^{(1)}(t_{k+1}) = \frac{3h_k (P_{k+1} - P_k)}{(3h_k + h_{k-1} + h_{k+1})}
\]

Hence from (2.18), (2.19), (2.20) if \(P(t) = P_k(t)\) for \(t \in (t_k, t_{k+1})\), then

\[
\lim_{\omega \to \infty} P_k(t) = (1 - \theta) \hat{F}_k + \theta \hat{F}_k.
\]
Proposition 2.6. (Local Tension) Consider an interval as in Proposition 5. Then on $[t_k, t_{k+1}]$, the weighted $\nu$-spline converges uniformly to the linear interpolant $l_i(t)$ as both $\nu_k, \nu_{k+1} \to \infty$, where $\omega_{k-1}, \omega_k, \omega_{k+1}$ are bounded.

Proof. It can be noted that
\[
\lim \mu_k = \lim \mu_{k+1} = 0,
\]
\[
\lim \lambda_k = \lim \lambda_{k+1} = 0,
\]
and
\[
\lim P^{(1)}(t_k) = \lim P^{(1)}(t_{k+1}) = 0.
\]
This gives the desired result.

2.3.4 Demonstration

The tension behavior of the weighted $\nu$-spline is illustrated by the following simple examples for data set in $\mathbb{R}^2$. Unless otherwise stated, in all the figures, the parameter $\nu_i$ will be assumed as zero $\forall i$ and the parameters $\omega_i$ as 1 for all $i$.

Figure 2.3 is the default curve, which is a cubic spline for $\nu_i = 0$, and $\omega_i = 1$, for all $i$. The control polygon, together with the control points, is also shown in the figure. Figure 2.4 shows the effect of a progressive increase in the interval tension in the base of the figure. The top, middle, and bottom curves have been demonstrated for $\omega = 1, 10,$ and 100, respectively. The effect of the high-tension parameters is clearly seen in the corresponding interval in the base of the figure. Figure 2.5 shows the effect of a progressive increase in point tension behavior locally at two opposite points of the figure. The top, middle, and bottom curves have been demonstrated for $\nu = 0, 10,$ and 100, respectively. The effect of the high-tension parameters is clearly seen at the corresponding points in the figure.
2.3. Freeform Weighted Nu Spline

Figure 2.4. The weighted Nu spline with interval tension at the base with \( \omega \) values as 1 (left curve), 10 (middle curve), 100 (right curve).

Figure 2.5. The weighted Nu spline with corner tension at two opposite points with \( \nu \) values as 0 (left curve), 10 (middle curve), 100 (right curve).

Figure 2.6. The weighted Nu spline with global tension \( \nu = 1 \) (top curve), \( \nu = 5 \) (middle curve), \( \nu = 100 \) (bottom curve).

Figures 2.6 illustrates the effect of progressively increasing the values of the point tension parameters \( \nu_i \)'s = 0, 5, and 100, for the top, middle, and bottom curves, respectively, at all the points of the figure. This is the global tension effect due to progressive increase.

Figure 2.7 demonstrates an important observation about the negative values of the shape parameters. The global values of the interval shape parameters \( \omega \)'s will not make any effect to the picture. However, the local values do influence the picture. The curve bulges inside for negative values \( \omega = 0, -3, -4, -5, -25, \) and \(-100\). It can be noted (row-wise from left to right) that lower negative values make the curve bulge more inside, but higher negative values again start making the curve tensed in the interval.

Behavior of the negative \( \nu \) values can be seen in Figure 2.8. It illustrates the effect of progressive negative increase in the values of the point tension parameters \( \nu_i \)'s = 0, \(-1, -5, -25, -100\), and \(-1000\). It can be seen (row-wise from left to right) that lower negative values make the curve bulge inside so much so the curve starts looping with the negative increase. However, it again starts getting tensed after attaining certain values. Ultimately, higher negative values make the curve tensed to converge to the control polygon.
Figure 2.7. The weighted Nu spline curves (row vise from left to right) with negative global tension $\omega$ values $0, -3, -4, -5, -25, 100$.

Figure 2.8. The weighted Nu spline curves (row vise from left to right) with negative global tension $\nu$ values $0, -1, -5, -25, 100, and 1000$. 
2.3.5 Advantages and Features

The method has various advantages and features as follows:

- It enjoys the good features of cubic splines.
- It enjoys all the standard geometric properties of B-splines.
- The method is geometrically smooth.
- It recovers the cubic B-spline method as a special case.
- It recovers the weighted spline method as a special case.
- It recovers the Nu-spline method as a special case.
- It possesses interested shape design features.
- It has two families of shape parameters working in such a way that one family of parameters is associated with intervals and the other with points. These parameters provide a variety of shape controls such as point and interval tension.
- Negative weights can also be utilized for shape design.
- It is computationally economical because it consumes the cubic function only.
- The method of evaluation is suggested by a transformation to Bézier form, which is computable by any well-known recursive method too.
- In addition to direct manipulation, the interpolation method can be computed through B-spline-like formulation too. This point will be discussed in detail somewhere else later.
- The curve method is extendable to surfaces. The direct approach using a tensor product is the simplest one.

2.4 Surfaces

The extension of the curve scheme, to tensor product surface representations:

\[ P(\tilde{t}, t) = \sum_{i=-1}^{m+1} \sum_{j=-1}^{n+1} P_{i,j} \tilde{B}_i(\tilde{t}) B_j(t), \]

where \( \tilde{t}_{m+3} \leq \tilde{t} \leq \tilde{t}_{m+3}, t_{n+3} \leq t \leq t_{n+3} \), is immediately apparent. This surface presents a bicubic weighted \( \nu \)-spline surface with shape parameters as:

- \( \tilde{\nu}_i \geq 0, \ i = 1, \ldots, m, \tilde{\omega}_j > 0, \ i = 1, \ldots, m - 1, \)
- \( \nu_j \geq 0, \ j = 1, \ldots, n, \omega_j > 0, \ j = 1, \ldots, n - 1. \)

Here \( P_{i,j} \in \mathbb{R}^3, i = -1, \ldots, m + 1, \ j = -1, \ldots, n + 1. \) are the data points and \( \tilde{B}_i, i = -1, \ldots, m + 1 \) and \( B_j, j = -1, \ldots, n + 1 \) are the local support bases functions for the weighted \( \nu \)-spline in \( \tilde{t} \) and \( t \) directions, respectively. However, this representation exhibits a problem common to all tensor product descriptions in that the shape control parameters now affect a complete row or column of the tensor product array.
2. Weighted Nu Splines

Nielsen [12] solves this problem for his cubic ν-spline representation by constructing a Boolean sum, spline-blended, rectangular network of parametric ν-spline curves. Another possibility is to allow the shape parameters to be variable in the orthogonal direction to, for example, the local support basis functions of the tensor product form.

We propose a tensor product like the approach in [16, 17], but actually it is not a tensor product. Instead of step functions, the tension weights are introduced as $C^2$ continuous cubic B-splines in the description of the tensor product. This produces local control in the construction of surfaces in an independent way. The details of the proposed method are out of the scope of this paper and will be discussed elsewhere.

2.5 Summary

A freeform $C^1$ weighted Nu spline curve design has been developed through the construction of local support B-spline-like basis functions. This cubic spline method has been developed with a view to its application in computer graphics, geometric modeling, and CAGD. It is quite reasonable to construct a freeform cubic spline method, which involves two families of shape parameters in exactly a similar way as in interpolatory weighted ν-spline. These parameters provide a variety of local and global shape controls such as interval and point shape effects. The visual smoothness of the proposed method is also $C^1$, which is same as the smoothness of interpolatory weighted ν-spline. The freeform $C^1$ weighted Nu-spline method can be applied to tensor product surfaces, but unfortunately, in the context of interactive surface design, this tensor product surface is not that useful because any one of the tension parameters controls an entire corresponding interval strip of the surface. Thus, as an application of $C^1$ spline for the surfaces, a method similar to Nielsen’s [12] spline blended methods may be attempted. This will produce local shape control, which is quite useful regarding the computer graphics and geometric modeling applications.

2.6 Exercises

1. Write a program to implement the curve design method in Section 2.2.
2. Write a program to implement the curve design method in Section 2.3.
3. Check the difference of shape effects in your programs of Exercise 2.6.1 and 2.6.2 when the schemes are implemented in scalar form.

References

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