

SPLINE COALESCENCE HIDDEN VARIABLE FRACTAL INTERPOLATION FUNCTIONS

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This paper generalizes the classical spline using a new construction of spline coalescence hidden variable fractal interpolation function (CHFIF). The derivative of a spline CHFIF is a typical fractal function that is self-affine or non-self-affine depending on the parameters of a nondiagonal iterated function system. Our construction generalizes the construction of Barnsley and Harrington (1989), when the construction is not restricted to a particular type of boundary conditions. Spline CHFIFs are likely to be potentially useful in approximation theory due to effects of the hidden variables and these effects are demonstrated through suitable examples in the present work.

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1. Introduction

The class of self-affine attractors was introduced by Mandelbrot [12], and its construction through iterated function systems (IFSs) was proposed by Hutchinson [11]. By using IFSs, Barnsley [1, 2] and Barnsley et al. [3] introduced the concept of fractal interpolation functions (FIFs), that are used for approximating naturally occurring functions, for example, speech signals [13], seismic data [10], electrocardiograms [15], images [5], which show some kind of geometrical self-similarity under magnification. For instance, in a case of a speech signal, the approximation is achieved by replacing the original data with the set of generating parameters (which are actually matrix elements of certain contractive affine mappings) used for generating a suitable FIF. The non-self-affine functions are approximated by using hidden variable FIFs studied by Barnsley [2], Barnsley et al. [3], and Massopust [14]. In [7], the projection of the graphs of vector-valued functions or attractors from a nondiagonal IFS is used to approximate self-affine or

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non-self-affine functions simultaneously by introducing the concept of *constrained free variables*. The smoothness analysis of CHFIF [6] gives that the deterministic construction of functions having modulus of continuity $O(|t|^\mu(\log|t|)^m)$ (m is an integer, $m \geq 0$, and $0 < \mu \leq 1$) is possible.

The construction of self-affine spline FIF with only initial endpoint boundary conditions is given by Barnsley and Harrington [4]. However, generating spline FIFs with general boundary condition is not possible in their construction due to a particular recurrence relation between the matrices. Some results on the convergence of cubic spline FIFs on uniform meshes are found in [17]. The general construction of spline FIFs and their convergence results are developed recently in [6, 8, 9, 16–19]. A CHFIF can be integrated successively in order to get smoother interpolation functions that generalize classical splines. The resulting smooth functions are also attractors of appropriate nondiagonal IFSs but they interpolate different data in general. The inverse problem is the construction of a spline CHFIF that interpolates the given data. Although spline CHFIFs are not actually fractals, certain derivatives of these functions are typical fractal functions that may be self-affine or non-self-affine in nature. The fractal dimension of a suitable derivative of spline CHFIFs can be aptly used as a quantifying parameter for various complex phenomena.

In the present paper, the existence of interpolating C^r -CHFIF for given real data with all possible boundary conditions is established by studying the calculus of vector-valued C^1 -FIF. The functional relations of C^r -CHFIF at the endpoints of the interval with join-up conditions and interpolation conditions give a system of equations whose solution determines the polynomial coefficients in the construction of a nondiagonal IFS. The advantage of such a construction is that with suitable choices of the hidden variables for prescribed data and given boundary conditions, one can construct self-affine or non-self-affine spline function according to the need of an experiment from a number of available spline CHFIF.

The organization of this paper is as follows: in Section 2, preliminaries concerning CHFIF are discussed. The basic calculus of vector-valued C^1 -FIF is studied in Section 3. The spline CHFIF with general boundary conditions is constructed in Section 4. The effects of hidden variables on spline CHFIFs are demonstrated in Section 5 through various examples.

2. Preliminaries of CHFIF

Suppose that the real data to be interpolated is given by $\{(x_n, y_n) \in \mathbb{R}^2 : n = 0, 1, 2, \dots, N\}$, where $-\infty < x_0 < x_1 < \dots < x_N < \infty$. In order to construct an interpolation function $f_1 : [x_0, x_N] \rightarrow \mathbb{R}$ such that $f_1(x_n) = y_n$, for all $n = 0, 1, 2, \dots, N$, consider a generalized set of data $\{(x_n, y_n, z_n) \in \mathbb{R}^3 \mid n = 0, 1, 2, \dots, N\}$, where the z_n , $n = 0, 1, 2, \dots, N$, are finite real numbers. The following notation is needed in the construction of CHFIF: $I = [x_0, x_N]$, $I_n = [x_{n-1}, x_n]$, $g_1 = \text{Min}_n y_n$, $g_2 = \text{Max}_n y_n$, $h_1 = \text{Min}_n z_n$, $h_2 = \text{Max}_n z_n$, and $K = I \times D$, where $D = J_1 \times J_2$, J_1, J_2 are suitable compact sets in \mathbb{R} such that $[g_1, g_2] \times [h_1, h_2] \subset D$. Let $L_n : I \rightarrow I_n$ be a contractive homeomorphism and let $F_n : K \rightarrow D$ be a continuous

vector-valued function such that

$$\begin{aligned} L_n(x_0) &= x_{n-1}, & L_n(x_N) &= x_n, \\ F_n(x_0, y_0, z_0) &= (y_{n-1}, z_{n-1}), & F_n(x_N, y_N, z_N) &= (y_n, z_n), \end{aligned} \tag{2.1}$$

$$\begin{aligned} d(F_n(x, y, z), F_n(x^*, y, z)) &\leq c|x - x^*|, \\ d(F_n(x, y, z), F_n(x, y^*, z^*)) &\leq sd_E((y, z), (y^*, z^*)), \end{aligned} \tag{2.2}$$

for all $n = 1, 2, \dots, N$, where c and s are positive constants with $0 \leq s < 1$, $(x, y, z), (x^*, y, z), (x, y^*, z^*) \in K$, d is the sup norm on K , and d_E is the Euclidean metric on \mathbb{R}^2 . For defining the required CHFIF, the functions L_n and F_n are chosen to be of the form $L_n(x) = a_n x + b_n$ and

$$F_n(x, y, z) = A_n(y, z)^T + (p_n(x), q_n(x))^T = (F_n^1(x, y, z), F_n^2(x, z))^T, \tag{2.3}$$

where A_n is an upper triangular matrix $\begin{pmatrix} \alpha_n & \beta_n \\ 0 & \gamma_n \end{pmatrix}$ and $p_n(x), q_n(x)$ are continuous functions having at least two unknowns. We choose α_n as free variable with $|\alpha_n| < 1$ and β_n as *constrained free variable* with respect to γ_n such that $|\beta_n| + |\gamma_n| < 1$. The generalized IFS that is needed for construction of CHFIF corresponding to the data $\{(x_n, y_n, z_n) \mid n = 0, 1, \dots, N\}$ is now defined as

$$\{\mathbb{R}^3; \omega_n(x, y, z) = (L_n(x), F_n(x, y, z)), n = 1, 2, \dots, N\}. \tag{2.4}$$

It is known [3] that the IFS defined in (2.4) associated with the data $\{(x_n, y_n, z_n), n = 0, 1, \dots, N\}$ is hyperbolic with respect to a metric ρ on \mathbb{R}^3 equivalent to the Euclidean metric. In particular, there exists a unique nonempty compact set $G \subseteq \mathbb{R}^3$ such that

$$G = \bigcup_{i=1}^N \omega_i(G). \tag{2.5}$$

The existence of a unique vector-valued function f that interpolates the generalized interpolation data by establishing the graph of f that equals the attractor G of the generalized IFS is given in the following proposition.

PROPOSITION 2.1 [3]. *The above attractor G of the IFS defined in (2.4) is the graph of the continuous vector-valued function $f : I \rightarrow D$ such that $f(x_n) = (y_n, z_n)$ for all $n = 1, 2, \dots, N$, that is, $G = \{(x, y, z) : x \in I \text{ and } f(x) = (y(x), z(x))\}$.*

Let the vector-valued function $f : I \rightarrow D$ in Proposition 2.1 be written as $f(x) = (f_1(x), f_2(x))$. The required CHFIF is now defined as follows.

Definition 2.2. Suppose $\{(x, f_1(x)) : x \in I\}$ is the projection of the attractor G (cf. (2.5)) on \mathbb{R}^2 . Then, the function $f_1(x)$ is called a *coalescence hidden variable FIF* (CHFIF) for the given interpolation data $\{(x_n, y_n) \mid n = 0, 1, \dots, N\}$.

Proposition 2.1 gives that the graph of the vector-valued function $f(x)$ is the attractor of the IFS given by (2.4) if and only if the fixed point f of the Read-Bajraktarević operator

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T on the space of continuous vector-valued functions from I to D satisfies

$$Tf(x) = f(x) = F_n(L_n^{-1}(x), f(L_n^{-1}(x))), \quad x \in I_n, n = 1, 2, \dots, N. \quad (2.6)$$

The image Tf of the vector-valued function f can be written componentwise as $(T_1 f_1, T_2 f_2)$, where T_1 and T_2 are the componentwise Read-Bajraktarević operators from I to \mathbb{R} . Thus, CHFIFs satisfy the following functional equation for all $x \in I$:

$$T_1 f_1(L_n(x)) = f_1(L_n(x)) = F_n^1(x, f_1(x), f_2(x)) = \alpha_n f_1(x) + \beta_n f_2(x) + p_n(x). \quad (2.7)$$

Similarly, the projection $\{(x, f_2(x)) : x \in I\}$ of the attractor G is self-affine in nature. The fractal function $f_2(x)$ interpolates the data $\{(x_n, z_n) : n = 0, 1, \dots, N\}$ and satisfies the functional equation for all $x \in I$,

$$T_2 f_2(L_n(x)) = f_2(L_n(x)) = F_n^2(x, f_2(x)) = \gamma_n f_2(x) + q_n(x). \quad (2.8)$$

Since the graph of f_1 is a projection of G , it need not be union of its affine transformations. Hence, CHFIFs are generally non-self-affine in nature. If $y_n = z_n$ and $\alpha_n + \beta_n = \gamma_n$, CHFIF $f_1(x)$ coincides with the self-affine fractal function $f_2(x)$ for the same interpolation data. Hence, the CHFIF from a nondiagonal IFS is self-affine in this case. Also, if $\beta_n = 0$, for $n = 1, 2, \dots, N$, then the CHFIF f_1 is obviously always self-affine. It is observed that [9] depending on the parameter of nondiagonal IFS (cf. (2.4)), the modulus of continuity of f_1 is $O(|t|^\mu (\log t)^\mu)$, where $n = 0, 1$, or 2 .

3. Calculus of vector-valued C^1 -FIF

First, it is shown that the integral of vector-valued FIF f is also a vector-valued FIF to a different set of interpolation data if any of the values (\hat{y}_0, \hat{z}_0) or (\hat{y}_N, \hat{z}_N) of the integral of f is known at an endpoint of the interval. Set $\hat{y}_N = \hat{y}_0 + (\sum_{i=1}^N a_i \{\beta_i (\hat{z}_N - \hat{z}_0) + \int_{x_0}^{x_N} p_i(\tau) d\tau\}) / (1 - \sum_{i=1}^N a_i \alpha_i)$, $\hat{z}_N = \hat{z}_0 + (\sum_{i=1}^N a_i \int_{x_0}^{x_N} q_i(\tau) d\tau) / (1 - \sum_{i=1}^N a_i \gamma_i)$, $\hat{p}_n(x) = \hat{y}_{n-1} - a_n(\alpha_n \hat{y}_0 + \beta_n \hat{z}_0) + a_n \int_{x_0}^x p_n(\tau) d\tau$, $\hat{q}_n(x) = \hat{z}_{n-1} - a_n \gamma_n \hat{z}_0 + a_n \int_{x_0}^x q_n(\tau) d\tau$, and

$$\hat{y}_n = \hat{y}_0 + \sum_{i=1}^n a_i \left\{ \alpha_i (\hat{y}_N - \hat{y}_0) + \beta_i (\hat{z}_N - \hat{z}_0) + \int_{x_0}^{x_N} p_i(\tau) d\tau \right\}, \quad (3.1)$$

$$\hat{z}_n = \hat{z}_0 + \sum_{i=1}^n a_i \left\{ \gamma_i (\hat{z}_N - \hat{z}_0) + \int_{x_0}^{x_N} q_i(\tau) d\tau \right\}, \quad (3.2)$$

$$\hat{F}_n(x, y, z) = (\hat{F}_n^1(x, y, z), \hat{F}_n^2(x, z)), \quad (3.3)$$

where $\hat{F}_n^1(x, y, z) = a_n(\alpha_n y + \beta_n z) + \hat{p}_n(x)$, $\hat{F}_n^2(x, z) = a_n \gamma_n z + \hat{q}_n(x)$.

PROPOSITION 3.1. *Let f be the vector-valued FIF associated with $\{(L_n(x), F_n(x, y, z)); n = 1, 2, \dots, N\}$, where F_n is defined by (2.3). Let*

$$\hat{f}_1(x) = \hat{y}_0 + \int_{x_0}^x f_1(\tau) d\tau, \quad \hat{f}_2(x) = \hat{z}_0 + \int_{x_0}^x f_2(\tau) d\tau, \quad (3.4)$$

then $\hat{f} = (\hat{f}_1, \hat{f}_2)$ is the vector-valued FIF associated with $\{(L_n(x), \hat{F}_n(x, y, z)); n = 1, 2, \dots, N\}$ such that $\hat{f}(x_n) = (\hat{y}_n, \hat{z}_n)$ for $n = 0, 1, 2, \dots, N$, where \hat{y}_n , \hat{z}_n , and $\hat{F}_n^1(x, y, z)$ are defined by (3.1)–(3.3), respectively.

Proof. Using the definition of \hat{f}_1 for $x \in I$,

$$\begin{aligned} \hat{f}_1(L_n(x)) &= \hat{y}_0 + \int_{x_0}^{L_n(x)} f_1(\tau) d\tau = \hat{y}_0 + \int_{x_0}^{x_{n-1}} f_1(\tau) d\tau + \int_{x_{n-1}}^{L_n(x)} f_1(\tau) d\tau \\ &= \hat{f}_1(x_{n-1}) + a_n \int_{x_0}^x f_1(L_n(\tau)) d\tau. \end{aligned} \tag{3.5}$$

So, using (2.7),

$$\begin{aligned} \hat{f}_1(L_n(x)) &= \hat{y}_{n-1} + a_n \int_{x_0}^x (\alpha_n f_1(\tau) + \beta_n f_2(\tau) + p_n(\tau)) d\tau \\ &= \hat{y}_{n-1} + a_n [\alpha_n (\hat{f}_1(x) - \hat{y}_0) + \beta_n (\hat{f}_2(x) - \hat{z}_0)] + a_n \int_{x_0}^x p_n(\tau) d\tau. \end{aligned} \tag{3.6}$$

In view of (2.7) and (3.6), the functional equation for \hat{f}_1 can be defined as follows:

$$\hat{F}_n^1(x, y, z) = \hat{f}_1(L_n(x)) = a_n (\alpha_n y + \beta_n z) + \hat{p}_n(x). \tag{3.7}$$

Similarly, using (2.8), the functional equation for \hat{f}_2 can be defined as

$$\hat{F}_n^2(x, y, z) = \hat{f}_2(L_n(x)) = a_n \gamma_n z + \hat{q}_n(x). \tag{3.8}$$

Thus, \hat{f}_1 is the CHFIF associated with $\{(L_n(x), \hat{F}_n(x, y, z)); n = 1, 2, \dots, N\}$ to a different set of interpolation data. Since \hat{f}_1 and \hat{f}_2 are continuous functions, the continuity conditions are valid at the new interpolation points.

The functional values for the CHFIF \hat{f}_1 are obtained by putting $x = x_N$ in (3.6). Hence,

$$\hat{y}_n = \hat{y}_{n-1} + a_n [\alpha_n (\hat{y}_N - \hat{y}_0) + \beta_n (\hat{z}_N - \hat{z}_0)] + a_n \int_{x_0}^{x_N} p_n(\tau) d\tau. \tag{3.9}$$

Inductively, using this equation up to $n = 1$, (3.1) follows. Putting $n = N$ in (3.1) and solving for \hat{y}_N give the relation between \hat{y}_N and \hat{y}_0 . Similarly, the functional values for the FIF \hat{f}_2 can be obtained by simplifying the expression for $\hat{f}_2(L_n(x))$ and are omitted. \square

Remark 3.2. If the value of integral of the vector-valued FIF is known at the final endpoint x_N instead of the initial endpoint, an analogue of Proposition 3.1 can be found as follows. Define

$$\hat{f}_1(x) = \hat{y}_N - \int_x^{x_N} f_1(\tau) d\tau, \quad \hat{f}_2(x) = \hat{z}_N - \int_x^{x_N} f_2(\tau) d\tau, \tag{3.10}$$

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then $\hat{f} = (\hat{f}_1, \hat{f}_2)$ is the vector-valued FIF associated with $\{(L_n(x), \hat{F}_n(x, y, z)); n = 1, 2, \dots, N\}$, where $\hat{F}_n(x, y, z) = (\hat{F}_n^1(x, y, z), \hat{F}_n^2(x, z))$, $\hat{F}_n^1(x, y, z) = a_n(\alpha_n y + \beta_n z) + \hat{p}_n(x)$, $\hat{F}_n^2(x, z) = a_n \gamma_n z + \hat{q}_n(x)$, $\hat{p}_n(x) = \hat{y}_n - a_n(\alpha_n \hat{y}_N + \beta_n \hat{z}_N) - a_n \int_{x_0}^{x_N} p_n(\tau) d\tau$, $\hat{y}_n = \hat{y}_N + \sum_{i=n+1}^N a_i \{\alpha_i (\hat{y}_N - \hat{y}_0) + \beta_i (\hat{z}_N - \hat{z}_0) + \int_{x_0}^{x_N} p_i(\tau) d\tau\}$, $\hat{q}_n(x) = \hat{z}_n - a_n \alpha_n \hat{z}_N - a_n \int_{x_0}^{x_N} q_n(\tau) d\tau$, $\hat{z}_n = \hat{z}_N - \sum_{i=n+1}^N a_i \{\alpha_i (\hat{z}_N - \hat{z}_0) + \int_{x_0}^{x_N} q_i(\tau) d\tau\}$, \hat{y}_0 and \hat{y}_N , \hat{z}_0 and \hat{z}_N are related in same manner as in Proposition 3.1.

Using Proposition 3.1 and Remark 3.2, a relation between a vector-valued FIF and its primitive is found through their IFS in the following proposition.

PROPOSITION 3.3. *Let f and \hat{f} be the vector-valued FIFs defined in Proposition 3.1 or Remark 3.2. Then, f is a primitive of \hat{f} if and only if \hat{f} is the vector-valued FIF associated with $\{(L_n(x), \hat{F}_n(x, y, z)); n = 1, 2, \dots, N\}$, where for $n = 1, 2, \dots, N$,*

$$\begin{aligned} \hat{F}_n(x, y, z) &= (\hat{F}_n^1(x, y, z), \hat{F}_n^2(x, z)), \\ \hat{F}_n^1(x, y, z) &= \hat{\alpha}_n y + \hat{\beta}_n z + \hat{p}_n(x), & \hat{F}_n^2(x, z) &= \hat{\gamma}_n z + \hat{q}_n(x), \\ \frac{\hat{\alpha}_n}{a_n} &= \alpha_n, & \frac{\hat{\beta}_n}{a_n} &= \beta_n, & \frac{\hat{\gamma}_n}{a_n} &= \gamma_n, & \hat{p}'_n &= a_n p_n, & \hat{q}'_n &= a_n q_n. \end{aligned} \quad (3.11)$$

Proof. The necessary part is a direct consequence of Proposition 3.1 and Remark 3.2 by using the fundamental theorem of Calculus. For sufficiency, the join-up conditions of the IFS give the following relations:

$$\begin{aligned} \hat{y}_n - \hat{y}_{n-1} &= \hat{F}_n^1(x_N, \hat{y}_N, \hat{z}_N) - \hat{F}_n^1(x_0, \hat{y}_0, \hat{z}_0) \\ &= \hat{\alpha}_n (\hat{y}_N - \hat{y}_0) + \hat{\beta}_n (\hat{z}_N - \hat{z}_0) + \hat{p}_n(x_N) - \hat{p}_n(x_0) \\ &= a_n [\alpha_n (\hat{y}_N - \hat{y}_0) + \beta_n (\hat{z}_N - \hat{z}_0)] + a_n \int_{x_0}^{x_N} p_n(\tau) d\tau. \end{aligned} \quad (3.12)$$

Since both values of \hat{y}_N and \hat{y}_0 are known if any one of these is given, the above relation inductively generates the interpolation values \hat{y}_n of \hat{f}_1 as $\hat{y}_n = \hat{y}_0 - \sum_{i=1}^n (\hat{y}_i - \hat{y}_{i-1}) = \hat{y}_0 - \sum_{i=1}^n a_i \{\alpha_i (\hat{y}_N - \hat{y}_0) + \beta_i (\hat{z}_N - \hat{z}_0) + \int_{x_0}^{x_N} p_i(\tau) d\tau\}$, $n = 1, \dots, N-1$.

Further, the condition $\hat{p}'_n = a_n p_n$ gives $\hat{p}_n(x) = a_n \int_{x_0}^x p_n(\tau) d\tau + H_n$ for some constant H_n . The constant H_n , uniquely determined by the join-up condition $\hat{F}_n(x_0, \hat{y}_0, \hat{z}_0) = (\hat{y}_{n-1}, \hat{z}_{n-1})$, is given by $H_n = \hat{y}_{n-1} - a_n \alpha_n \hat{y}_0 - a_n \beta_n \hat{z}_0$. Thus, \hat{p}_n is also uniquely determined as $\hat{p}_n(x) = \hat{y}_{n-1} - a_n (\alpha_n \hat{y}_0 + \beta_n \hat{z}_0) + a_n \int_{x_0}^x p_n(\tau) d\tau$. Since the expressions for \hat{y}_n and \hat{p}_n are the same as in Proposition 3.1, it follows that $\hat{f}'_1 = f_1$. Similarly, using the relation between \hat{z}_n and \hat{z}_{n-1} , the join-up condition, and Proposition 3.1, $\hat{f}'_2 = f_2$. Analogously, if \hat{f} is defined as in Remark 3.2, it can be seen that $\hat{f}' = f$. \square

Remark 3.4. In the construction of \hat{f} from the given vector-valued FIF f obtained in Proposition 3.1 and Remark 3.2, the value of \hat{f} is assumed to be known at one of the

endpoints of the interval I . Hence, using Proposition 3.1 and Remark 3.2, a twice differentiable vector-valued FIF can be constructed when values of \hat{f} and its integral are known at any combination of the endpoints of the interval I . In general, using Proposition 3.1 and Remark 3.2, we can construct a vector-valued C^r -FIF that interpolates a different data, when the values of each of r successive integrals of a vector-valued FIF are prescribed at any combination of endpoints.

4. Construction of spline CHFIF

The construction of spline CHFIF or C^r -CHFIF that interpolates the prescribed data is found as the fixed point of a suitably chosen IFS and the parameters of such an IFS are determined by finding the solution of a system of equations in which any type of boundary conditions can be considered.

Let $\{(x_0, y_0), (x_1, y_1), \dots, (x_N, y_N)\}$, $x_0 < x_1 < \dots < x_N$, be the given data points and it is extended to generalized data by providing the real parameters z_0, z_1, \dots, z_N . Let $\mathcal{G}^r = \{g \in C^r(I, \mathbb{R}^2) \mid g(x_0) = (y_0, z_0) \text{ and } g(x_N) = (y_N, z_N)\}$, where r is some nonnegative integer. Let d^* be the C^r -norm on \mathcal{G}^r . For $x \in I_n$, $n = 1, 2, \dots, N$, define the Read-Bajraktarević operator T on the space (\mathcal{G}^r, d^*) of vector-valued functions by

$$\begin{aligned} Tg(x) &= F_n(L_n^{-1}(x), g(L_n^{-1}(x))) \\ &= (\alpha_n g_1(L_n^{-1}(x)) + \beta_n g_2(L_n^{-1}(x)) + p_n(x), \gamma_n g_2(L_n^{-1}(x)) + q_n(x)), \end{aligned} \tag{4.1}$$

where $L_n(x) = a_n x + b_n$ satisfies (2.1), $p_n(x), q_n(x)$ are suitably chosen polynomials, $|\alpha_n| < a_n^r$, and $|\beta_n| + |\gamma_n| < a_n^r$ for $n = 1, 2, \dots, N$. The restrictions on free variables and constrained free variables imply that T is a contractive operator on (\mathcal{G}^r, d^*) . Since (\mathcal{G}^r, d^*) is complete, the fixed point $f = (f_1, f_2)$ of T satisfies the functional relations, $f_1(L_n(x)) = \alpha_n f_1(x) + \beta_n f_2(x) + p_n(x)$ and $f_2(L_n(x)) = \gamma_n f_2(x) + q_n(x)$ for $n = 1, 2, \dots, N$. In general, f_1 interpolates data different from the prescribed data. Using Proposition 3.3, it follows that f_1' satisfies the functional relation

$$f_1'(L_n(x)) = \frac{\alpha_n f_1'(x) + \beta_n f_2'(x) + p_n'(x)}{a_n}, \quad n = 1, 2, \dots, N. \tag{4.2}$$

Since $|\alpha_n|/a_n \leq |\alpha_n|/a_n^r < 1$ and $(|\beta_n| + |\gamma_n|)/a_n \leq (|\beta_n| + |\gamma_n|)/a_n^r < 1$, f_1' is a CHFIF for different data. Inductively, using the above arguments, the following relations are obtained:

$$f_1^{(k)}(L_n(x)) = \frac{\alpha_n f_1^{(k)}(x) + \beta_n f_2^{(k)}(x) + p_n^{(k)}(x)}{a_n^k}, \quad n = 1, 2, \dots, N; \quad k = 0, 1, 2, \dots, r, \tag{4.3}$$

where $f_1^{(0)} = f_1, f_2^{(0)} = f_2$, and $p^{(0)} = p$. Since $|\alpha_n|/a_n^k \leq |\alpha_n|/a_n^r < 1$ and $(|\beta_n| + |\gamma_n|)/a_n \leq (|\beta_n| + |\gamma_n|)/a_n^r < 1$, $f_1^{(k)}$ ($k = 2, 3, \dots, r$) is a CHFIF for different data. In general, $f_1^{(k)}$ ($k = 2, 3, \dots, r$) interpolates data different from the prescribed data. In particular, $f_1^{(r)}$ is an affine CHFIF for different data, if the polynomial $p_n^{(r)}$ occurring in (4.3) with $k = r$

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is affine. Thus, $p_n(x)$ is to be chosen as a polynomial of degree $(r + 1)$ for f_1 to be a C^r -CHFIF. Let $p_n(x) = \sum_{k=0}^{r+1} p_{kn}x^k$, $n = 1, 2, \dots, N$, where the coefficients p_{kn} have to be determined suitably such that f_1 interpolates the prescribed data. The continuity of $f_1^{(k)}$ on I gives

$$f_1^{(k)}(L_{n+1}(x_0)) = f_1^{(k)}(L_n(x_N)), \quad k = 0, 1, \dots, r; \quad n = 1, 2, \dots, N - 1. \quad (4.4)$$

Hence, using (4.3), the following $(r + 1)(N - 1)$ join-up conditions can be obtained for $k = 0, 1, \dots, r$ and $n = 1, 2, \dots, N - 1$:

$$\frac{\alpha_{n+1}f_1^{(k)}(x_0) + \beta_{n+1}f_2^{(k)}(x_0) + p_{n+1}^{(k)}(x_0)}{a_{n+1}^k} = \frac{\alpha_n f_1^{(k)}(x_N) + \beta_n f_2^{(k)}(x_N) + p_n^{(k)}(x_N)}{a_n^k}. \quad (4.5)$$

Also, (4.3) implies that at the endpoints on the interval, the values of $f_1^{(k)}$ satisfy the following $2r$ -conditions:

$$\begin{aligned} f_1^{(k)}(x_0) &= \frac{\alpha_1 f_1^{(k)}(x_0) + \beta_1 f_2^{(k)}(x_0) + p_1^{(k)}(x_0)}{a_1^k}, \quad k = 1, 2, \dots, r, \\ f_1^{(k)}(x_N) &= \frac{\alpha_N f_1^{(k)}(x_N) + \beta_N f_2^{(k)}(x_N) + p_N^{(k)}(x_N)}{a_N^k}, \quad k = 1, 2, \dots, r. \end{aligned} \quad (4.6)$$

Finally, the given interpolation conditions are

$$f_1(x_n) = y_n, \quad n = 0, 1, \dots, N. \quad (4.7)$$

Similarly, if $q_n(x) = \sum_{k=0}^{r+1} q_{kn}x^k$, $n = 1, 2, \dots, N$, the analogous equations for the coefficients q_{kn} associated with self-affine fractal function f_2 are

$$\frac{\gamma_{n+1}f_2^{(k)}(x_0) + q_{n+1}^{(k)}(x_0)}{a_{n+1}^k} = \frac{\gamma_n f_2^{(k)}(x_N) + q_n^{(k)}(x_N)}{a_n^k}, \quad k = 0, 1, \dots, r; \quad n = 1, 2, \dots, N - 1,$$

$$f_2^{(k)}(x_0) = \frac{\gamma_1 f_2^{(k)}(x_0) + q_1^{(k)}(x_0)}{a_1^k}, \quad k = 1, 2, \dots, r,$$

$$f_2^{(k)}(x_N) = \frac{\gamma_N f_2^{(k)}(x_N) + q_N^{(k)}(x_N)}{a_N^k}, \quad k = 1, 2, \dots, r,$$

$$f_2(x_n) = z_n, \quad n = 0, 1, \dots, N. \quad (4.8)$$

The set of (4.8) involves $(r + 1)(N - 1) + 2r + (N + 1) = (r + 2)N + r$ conditions and $(r + 2)N + 2r$ number of unknowns (i.e., $f_2^{(k)}(x_0), f_2^{(k)}(x_N)$ for $k = 1, 2, \dots, r$ and the coefficients of the polynomials $q_n(x)$, q_{kn} , $k = 0, 1, \dots, r + 1$, $n = 1, 2, \dots, N$). Therefore, assuming r suitable relations involving the values of C^r -fractal function f_2 or the values of its derivatives at endpoints of $[x_0, x_N]$, the values of $f_2^{(k)}(x_0)$ and $f_2^{(k)}(x_N)$ for

$k = 1, 2, \dots, r$ can be obtained from (4.8). Now, from (4.5)–(4.7), the total number of the conditions for f_1 to interpolate the prescribed data is $(r + 2)N + r$. Equations (4.5)–(4.7) involve $2r$ unknowns: $f_1^{(k)}(x_0)$ and $f_1^{(k)}(x_N)$ for $k = 1, 2, \dots, r$ and $(r + 2)N$ unknown coefficients, p_{kn} , $k = 0, 1, \dots, r + 1$, $n = 1, 2, \dots, N$ of the polynomials $p_n(x)$. Consequently, in all $(r + 2)N + 2r$ number of unknowns are to be determined. The principle of construction of spline or C^r -CHFIF is to determine the above unknowns by choosing additional suitable r -restrictions on the values of C^r -CHFIF or the values of its derivatives at the boundary points of $[x_0, x_N]$ such that (4.5)–(4.7) together with these additional conditions are linearly independent. These unknowns are determined uniquely as solutions of these linearly independent systems of equations. The existence of the attractor for a C^r -CHFIF is guaranteed by the fixed point theorem and the assumptions $|\alpha_n| < a_n^r$ and $|\beta_n| + |\gamma_n| < a_n^r$. Hence, the required C^r -CHFIF f_1 interpolating the prescribed data is constructed as the attractor of the following IFS given by

$$\{\mathbb{R}^3; \omega_n(x, y) = (L_n(x), F_n(x, y, z)) = (\alpha_n y + \beta_n z + p_n(x), \gamma_n z + q_n(x)), n = 1, 2, \dots, N\}, \tag{4.9}$$

where $|\alpha_n| < a_n^r$, $|\beta_n| + |\gamma_n| < a_n^r$, $q_n(x)$, and $p_n(x)$, $n = 1, 2, \dots, N$, are the polynomials with coefficients q_{kn} and p_{kn} , respectively, computed by solving the linear independent system of equations, given by the above procedure. The flexibility of these choices of boundary conditions allows for the construction of a wide range of spline CHFIFs that may be self-affine or non-self-affine in nature. In such a construction, for a given choice of boundary conditions, depending upon the nature of the problem or simply the discretion of the user, an infinite number of suitable spline CHFIFs may be constructed due to the freedom of choices in the free variables α_n , γ_n , constrained free variables β_n for $n = 1, 2, \dots, N$, the free parameters z_n , $n = 0, 1, 2, \dots, N$, and the boundary conditions of the self-affine fractal function f_2 .

5. Examples of spline CHFIF

We computationally generate examples of spline CHFIF $f_1 \in C^2[0, 1]$ from the IFS given by (4.9). The interpolation data set is taken as $\{(0, 0), (2/5, 1), (3/4, -1), (1, 2)\}$ for spline CHFIF with $\alpha_n/a_n^2 = 0.8$, $n = 1, 2, 3$. The interpolation data is extended to generalized interpolation data with the addition of hidden variables $z_0 = 3$, $z_1 = 2$, $z_2 = 8$, $z_3 = 5$ and $\gamma_1/a_1^2 = 0.3$, $\gamma_2/a_2^2 = 0.35$, $\gamma_3/a_3^2 = 0.4$. Suppose that the constrained free variables are chosen as $\beta_1/a_1^2 = 0.4$, $\beta_2/a_2^2 = 0.6$, $\beta_3/a_3^2 = 0.5$ for a spline CHFIF (Figure 5.1). By using (2.1), $L_1(x) = (2/5)x$, $L_2(x) = (7/20x) + 2/5$, and $L_3(x) = (1/4x) + 3/4$ in the IFS for our examples.

First, we calculate the polynomial $q_n(x) = \sum_{k=0}^3 q_{kn}x^k$ for $n = 1, 2, 3$ for our first 4 examples. By the principle of construction of spline CHFIF, we choose $f_2'(x_0) = 10$ and $f_2'(x_3) = 1$ in order to solve system of (4.8). The computed $F_n^2(x, z)$ by using polynomials $q_n(x)$ are given in Table 5.1. For constructing the first example of spline CHFIF, we choose $f_1'(x_0) = 2$ and $f_1'(x_N) = 5$. With these choices, the system of (4.6)-(4.7) is solved to get the values of p_{kn} in $p_n(x) = \sum_{k=0}^3 p_{kn}x^k$ for $n = 1, 2, 3$. Using $F_n^1(x, y, z)$ (Table 5.2) in the IFS (4.9), the desired spline CHFIF is generated (Figure 5.1).

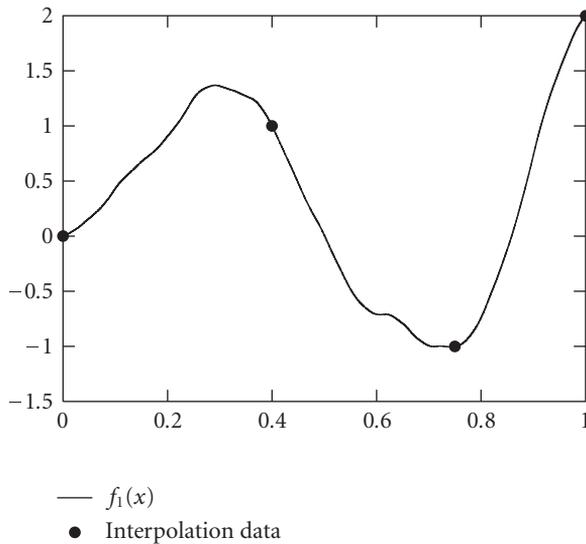


Figure 5.1. Spline CHFIF $f_1(x)$ for comparison.

Next, to construct an example of spline CHFIF with analogous boundary conditions as by Barnsley and Harrington [4], we choose $f_1'(x_0) = 2$ and $f_1''(x_0) = 5$. By the principle of construction of CHFIF, the coefficients of $p_n(x)$ are computed (Table 5.2). The iteration of the resulting IFS code (4.9) generates the spline CHFIF (Figure 5.2). Similarly, Figures 5.3 and 5.4 are generated to see the effect of change in free variables α_n and constrained free variables β_n , respectively, on the shape of spline CHFIF in comparison to that given by Figure 5.1 with the same choice of two additional conditions, $f_2'(x_0) = 10$ and $f_2'(x_3) = 1$. In Figure 5.3, we have chosen $\alpha_n/a_n^2 = -0.9$ for $n = 1, 3$ and $\alpha_2/a_2^2 = 0.9$. In Figure 5.4, we have chosen $\beta_1/a_1^2 = -0.6$, $\beta_2/a_2^2 = -0.6$ and $\beta_3/a_3^2 = 0.2$.

Next, we change the hidden variable γ_n as $\gamma_1/a_1^2 = -0.5$, $\gamma_2/a_2^2 = 0.3$, and $\gamma_3/a_3^2 = -0.4$. Subsequently, with the same choice of boundary conditions $f_2'(x_0) = 10$ and $f_2(x_3) = 1$, $F_n^2(x, z)$ is computed in Table 5.1. Using $f_2^{(k)}(x_0), f_2^{(k)}(x_3); k = 1, 2, F_n^1(x, y, z)$ (Table 5.2) is computed with $f_1'(x_0) = 2$ and $f_1'(x_3) = 5$ in order to see the effect of γ_n on the shape of spline CHFIF (Figure 5.5) in comparison to that given by Figure 5.1. In Figure 5.6, we change only the hidden variable z_n as $z_0 = -7, z_1 = -10, z_2 = 9, z_3 = -8$ in order to compare with Figure 5.1.

In Figure 5.7, we change the boundary condition of spline fractal function $f_2(x)$ as $f_2''(x_0) = 10, f_2''(x_3) = 1$. Consequently, the fractal function f_2 changes and the effect of change in boundary conditions of f_2 on the shape of spline CHFIF (Figure 5.7) can be seen by comparing with Figure 5.1.

In all seven spline CHFIFs above, the second derivative $f_1''(x)$ is a non-self-affine in nature. Finally, we assume $\gamma_n = z_n$ for $n = 0, 1, 2, 3, \alpha_n + \beta_n = \gamma_n$ for $n = 1, 2, 3$ with the same boundary conditions $f_1'(x_0) = f_2'(x_0) = 2$ and $f_1'(x_3) = f_2'(x_3) = 5$. Here, we have considered $\gamma_n/a_n^2 = 0.8$ for $n = 1, 2, 3$ and $\alpha_1/a_1^2 = 0.5, \alpha_2/a_2^2 = 0.4, \alpha_3/a_3^2 = 0.3, \beta_1/a_1^2 = 0.3,$

Table 5.1. $F_n^2(x, z)$ used in the construction of $f_1(x)$.

Figures 5.1–5.4	$F_1^2(x, z) = 0.048z + 7.1155x^3 - 11.7315x^2 + 3.52x + 2.856$ $F_2^2(x, z) = 0.0429z - 13.9198x^3 + 18.9927x^2 + 0.8413x + 1.8714$ $F_3^2(x, z) = 0.025z + 4.0109x^3 - 4.7468x^2 - 2.3141x + 7.925$
Figure 5.5	$F_1^2(x, z) = -0.08z + 11.882x^3 - 17.522x^2 + 4.8x + 3.24$ $F_2^2(x, y) = 0.0367z - 8.7332x^3 + 10.3705x^2 + 4.2892x + 1.8898$ $F_3^2(x, y) = -0.025z + 5.6159x^3 - 8.0068x^2 - 0.5591x + 8.075$
Figure 5.6	$F_1^2(x, z) = 0.048z + 16.9824x^3 - 23.4544x^2 + 3.52x - 6.664$ $F_2^2(x, z) = 0.0429z - 52.7327x^3 + 65.5487x^2 + 6.2269x - 9.6999$ $F_3^2(x, z) = 0.025z + 19.0457x^3 - 20.8914x^2 - 15.1293x + 9.175$
Figure 5.7	$F_1^2(x, z) = 0.048z + 3.6899x^3 - 0.7521x^2 - 4.0339x + 2.856$ $F_2^2(x, z) = 0.0429z - 5.9516x^3 + 7.7897x^2 + 4.0761x + 1.8714$ $F_3^2(x, z) = 0.025z + 1.5994x^3 - 5.2041x^2 + 0.5547x + 7.925$
Figure 5.8	$F_1^2(x, z) = 0.128z + 1.446x^3 - 1.246x^2 + 0.544x$ $F_2^2(x, z) = 0.098z + 11.83x^3 - 16.4813x^2 + 2.4552x + 1$ $F_3^2(x, z) = 0.05z - 0.9909x^3 + 0.0817x^2 + 3.8091x - 1$

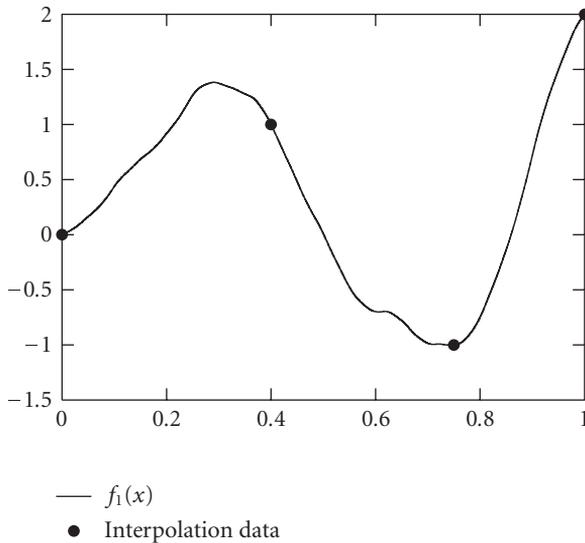


Figure 5.2. Effect on spline CHFIF by a change in boundary conditions of $f_1(x)$.

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Table 5.2. $F_n^1(x, y, z)$ used in the construction of $f_1(x)$.

Figure 5.1	$F_1^1(x, y, z) = 0.128y + 0.064z - 5.9811x^3 + 6.6931x^2 - 0.096x - 0.192$ $F_2^1(x, y, z) = 0.098y + 0.0735z - 0.1352x^3 + 2.1787x^2 - 4.3865x + 0.7795$ $F_3^1(x, y, z) = 0.05y + 0.0313z - 5.0268x^3 + 8.1848x^2 - 0.3205x - 1.0938$
Figure 5.2	$F_1^1(x, y, z) = 0.128y + 0.064z - 1.7189x^3 + 5.3755x^2 - 3.0406x - 0.192$ $F_2^1(x, y, z) = 0.098y + 0.0735z - 14.524x^3 + 21.7151x^2 - 9.5341x + 0.7795$ $F_3^1(x, y, z) = 0.05y + 0.0313z - 2.5515x^3 + 1.6161x^2 + 3.7729x - 1.0938$
Figure 5.3	$F_1^1(x, y, z) = -0.144y + 0.064z - 7.2999x^3 + 8.0119x^2 + 0.448x - 0.192$ $F_2^1(x, y, z) = 0.1102y + 0.0735z - 3.0304x^3 + 6.9418x^2 - 6.2789x + 0.7795$ $F_3^1(x, y, z) = -0.0563y + 0.0313z - 5.4156x^3 + 9.2811x^2 - 0.8156x - 1.0938$
Figure 5.4	$F_1^1(x, y, z) = 0.128y - 0.096z + 6.2655x^3 - 6.8335x^2 + 1.504x + 0.288$ $F_2^1(x, y, z) = 0.098y - 0.0735z + 16.2854x^3 - 25.1537x^2 + 6.8193x + 1.2205$ $F_3^1(x, y, z) = 0.05y + 0.0125z - 0.8557x^3 - 0.1762x^2 + 3.9068x - 1.0375$
Figure 5.5	$F_1^1(x, y, z) = 0.128y + 0.064z - 2.8759x^3 + 3.5879x^2 - 0.096x - 0.192$ $F_2^1(x, y, z) = 0.098y + 0.0735z + 5.0471x^3 - 5.7207x^2 - 1.6694x + 0.7795$ $F_3^1(x, y, z) = 0.05y + 0.0313z - 3.2659x^3 + 4.663x^2 + 1.4404x - 1.0938$
Figure 5.6	$F_1^1(x, y, z) = 0.128y + 0.064z - 27.2463x^3 + 28.1503x^2 - 0.096x + 0.448$ $F_2^1(x, y, z) = 0.098y + 0.0735z - 36.6824x^3 + 57.2174x^2 - 22.6575x + 1.5145$ $F_3^1(x, y, z) = 0.05y + 0.0313z - 17.9536x^3 + 33.9448x^2 - 13.0599x - 0.7813$
Figure 5.7	$F_1^1(x, y, z) = 0.128y + 0.064z + 1.0682x^3 - 1.8558x^2 + 1.4035x - 0.192$ $F_2^1(x, y, z) = 0.098y + 0.0735z + 10.1125x^3 - 13.3792x^2 + 0.9237x + 0.7795$ $F_3^1(x, y, z) = 0.05y + 0.0313z - 1.2489x^3 + 1.3367x^2 + 2.7497x - 1.0938$
Figure 5.8	$F_1^1(x, y, z) = 0.08y + 0.048z + 1.446x^3 - 1.246x^2 + 0.544x$ $F_2^1(x, y, z) = 0.049y + 0.049z + 11.83x^3 - 16.4813x^2 + 2.4552x + 1$ $F_3^1(x, y, z) = 0.0187y + 0.0313z - 0.9909x^3 + 0.0817x^2 + 3.8091x - 1$

$\beta_2/a_2^2 = 0.4$, $\beta_3/a_3^2 = 0.5$. In this case, the projection of nondiagonal IFS code generates self-affine spline CHFIF (Figure 5.8). Thus, our approach offers great flexibility and diversity to an experimenter depending upon the need of the problem for the choice of a self-affine or non-self-affine spline FIF.

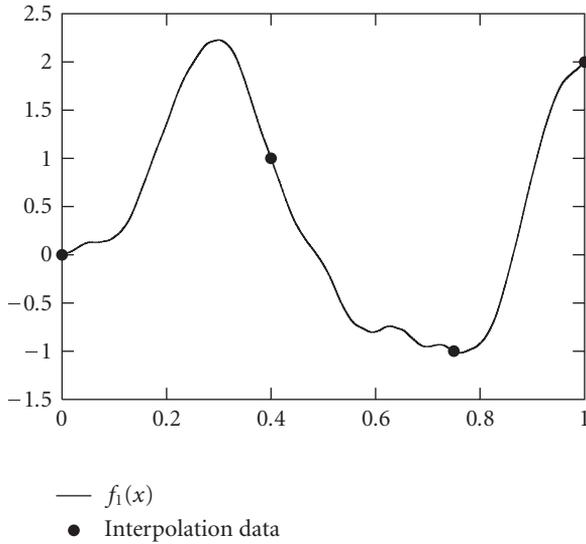


Figure 5.3. Effect on spline CHFIF by a change in the hidden variables α_n .

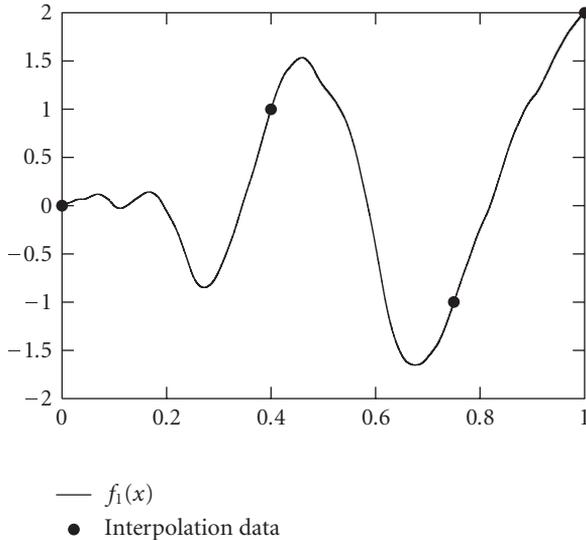


Figure 5.4. Effect on spline CHFIF by a change in the hidden variables β_n .

6. Conclusion

We have constructed the differentiable or spline CHFIFs for which certain derivative is a self-affine or non-self-affine fractal function depending on the nondiagonal IFS parameters. In our construction of a C^r -CHFIF f_1 , we chose $|\alpha_n| < a_n^r$ and $|\beta_n| + |\gamma_n| < a_n^r$ for

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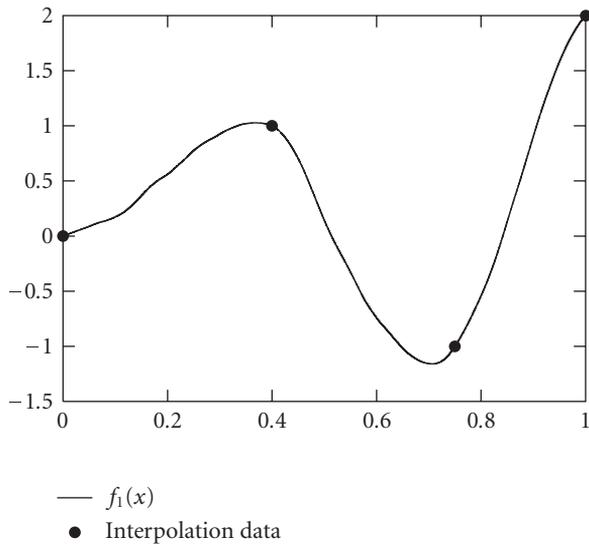


Figure 5.5. Effect on spline CHFIF by a change in the hidden variables γ_n .

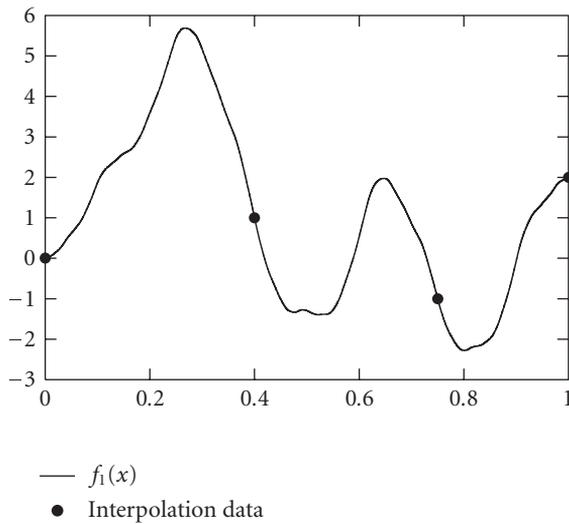


Figure 5.6. Effect on spline CHFIF by a change in the hidden variables z_n .

$n = 1, 2, \dots, N$, where a_n can be calculated from the given interpolation data. The system of (4.8) is solved with a suitable choice of boundary conditions for the coefficients of polynomial q_n and the derivatives of the fractal function f_2 at endpoints. Finally, by prescribing the boundary conditions for the CHFIF, the coefficients of polynomials p_n are

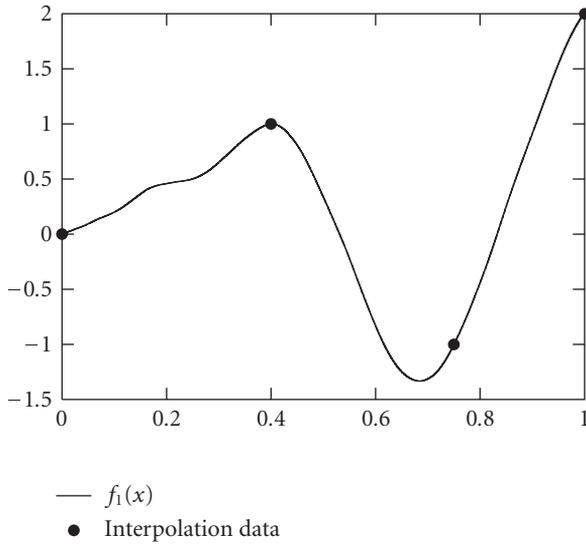


Figure 5.7. Effect on spline CHFIF by a change in the boundary conditions of $f_2(x)$.

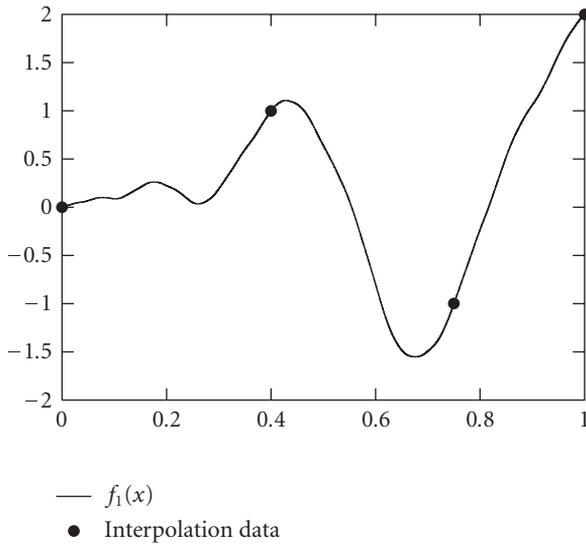


Figure 5.8. Self-affine spline CHFIF.

obtained from (4.5)–(4.7). The fixed-point theorem gives the existence of the attractor of the IFS given by (4.9) and the projection of the attractor is the desired C^r -CHFIF. Our construction allows admissibility of any kind of boundary conditions by using the solution of a system of equations. Our construction successfully answers the question “what

sort of endpoint condition leads to unique solution” of Barnsley’s and Harrington’s [4]. Thus, the complex algebraic method for construction of a spline FIF [4] by using complicated matrices with a particular type of boundary conditions is no longer needed. The hidden variables, free variables, and constrained variables play an important role in determining the shape of differentiable CHFIF that is very useful in approximation theory. For given boundary conditions, an infinite number of spline CHFIF can be constructed interpolating the same data by varying the free variables α_n , the constrained free variables β_n , the hidden variables γ_n , the free parameter z_n , or the *boundary conditions of self-affine spline fractal function*. It is felt that the spline CHFIF can be used in various scientific and engineering applications to capture the self-affine and non-self-affine nature simultaneously for relevant smooth objects.

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As a multidisciplinary field, financial engineering is becoming increasingly important in today's economic and financial world, especially in areas such as portfolio management, asset valuation and prediction, fraud detection, and credit risk management. For example, in a credit risk context, the recently approved Basel II guidelines advise financial institutions to build comprehensible credit risk models in order to optimize their capital allocation policy. Computational methods are being intensively studied and applied to improve the quality of the financial decisions that need to be made. Until now, computational methods and models are central to the analysis of economic and financial decisions.

However, more and more researchers have found that the financial environment is not ruled by mathematical distributions or statistical models. In such situations, some attempts have also been made to develop financial engineering models using intelligent computing approaches. For example, an artificial neural network (ANN) is a nonparametric estimation technique which does not make any distributional assumptions regarding the underlying asset. Instead, ANN approach develops a model using sets of unknown parameters and lets the optimization routine seek the best fitting parameters to obtain the desired results. The main aim of this special issue is not to merely illustrate the superior performance of a new intelligent computational method, but also to demonstrate how it can be used effectively in a financial engineering environment to improve and facilitate financial decision making. In this sense, the submissions should especially address how the results of estimated computational models (e.g., ANN, support vector machines, evolutionary algorithm, and fuzzy models) can be used to develop intelligent, easy-to-use, and/or comprehensible computational systems (e.g., decision support systems, agent-based system, and web-based systems)

This special issue will include (but not be limited to) the following topics:

- **Computational methods:** artificial intelligence, neural networks, evolutionary algorithms, fuzzy inference, hybrid learning, ensemble learning, cooperative learning, multiagent learning

- **Application fields:** asset valuation and prediction, asset allocation and portfolio selection, bankruptcy prediction, fraud detection, credit risk management
- **Implementation aspects:** decision support systems, expert systems, information systems, intelligent agents, web service, monitoring, deployment, implementation

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