

Shape Properties of Irregular Surface Data

Maria Hussain

Department of Mathematics

Lahore College for Women for University, Lahore, Pakistan

mariahussain_1@yahoo.com

Malik Zawwar Hussain

Department of Mathematics

University of the Punjab, Lahore, Pakistan

malikzawwar.math@pu.edu.pk

Iram Butt

Department of Mathematics

University of the Punjab, Lahore, Pakistan

Abstract

The presented work of this paper addresses the two shape properties, positivity and monotonicity of irregular surface data. The data is initially triangulated and a side-vertex scheme is adopted to interpolate the data over each triangle. Each boundary and radial curve is a rational function with three parameters facilitating 18 parameters in each triangular patch. The presence of these parameters leads to an automotive scheme for shape preservation and shape control. The data dependent constraints are derived on 6 of these parameters for preservation of positive and monotone properties of data, while, remaining 12 are free for shape modification. This scheme is local, does not constrain step length and derivatives, equally applicable to both data and data with derivatives.

Keywords: Rational cubic function, Triangular patch, Shape parameters, Positivity, Monotonicity.

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1. Introduction

The development of interpolation and approximation techniques which address the problem of shape preservation and control is one of the sought areas of Computer Graphics. The properties of data, whether gathered experimental or physical, can be categorized as positive, monotone and convex. Ordinary interpolation techniques guarantee smoothness but do not concentrate whether the shape properties of data are also inherited by developed curves and surfaces.

The data related to population distribution and probability distribution (Hussain et al., 1997) monthly rainfall amounts, progress of an irreversible process, level of gas discharged in certain chemical reaction, density and volume etc. (Hussain and Hussain, 2010) are always positive. A few examples of monotone data are dose-response curves and surfaces in biochemistry and pharmacology, approximation of couples and quasi

couples in statistics, empirical option pricing model in finance and approximation of potential functions in physical and chemical systems (Beliakov, 2005).

The limitation of ordinary interpolation techniques leads to the problem of shape preserving interpolation discussed by a number of authors. Chan and Ong (Chan and Ong, 2001) involved the cubic Bézier triangular interpolant for range restricted interpolation of irregular surface data. The derived sufficient conditions imposed restrictions on Bézier ordinates to give non-negative cubic Bézier triangular patch. The interpolating surface is a convex combination of these patches. If the Bézier ordinates did not satisfy the imposed restriction then it was remedied by scaling of first order partial derivatives at vertices. Piah, Goodman and Unsworth (Piah et al., 2005) also used cubic Bézier triangular interpolant to preserve the shape of irregular surface data. The derivatives at data sites were computed to be consistent with these conditions. Piah et al., (2006) addressed the problem of range restricted interpolation using quartic Bézier triangular interpolant which was an extension of (Piah et al., 2005). Each triangular patch was the convex combination of three quartic Bézier triangular patches. Again the positivity was ensured by a set of restriction on Bézier ordinates. Hussain and Hussain (2010, 2011) used rational function with parameters to preserve the positive and monotone shape of irregular surface data. Hussain and Hussain (2010, 2011) developed data dependent constraints on these parameters to preserve the positive and monotone shape of data and no parameter were free for shape refinement. Mulansky and Schmidt (1994) used quadratic spline on a Powell-Sabin refinement of triangulation to generate a constrained interpolant. Utreras (1985) defined a method that how positivity can be treated as a constraint, providing a global optimization at each step of iteration, but its computational cost was high.

Beliakov (2005) introduced a method for monotone interpolation and smoothing of irregular surface data. Monotonicity constraints were applied on noisy data to make it smooth and change it into a quadratic programming problem. This method was only useful to preserve the shape of monotone Lipschitz continuous function. Han and Schumaker (1997) scheme addressed the monotonicity of irregular surface data. The given irregular surface data was arranged over rectangular grids. The demerit of this method was that a system of N -irregular surface data points reduced to N^2 - rectangles. Moreover, a few of these were very small in one or both directions. Goodman et al., (1995) proposed derivative estimation scheme for scattered data.

The piecewise rational cubic function (Sarfraz, et al., 1997) having three shape parameters generates the radial and boundary curves of each triangular patch. Each triangular patch has six shape preserving parameters and twelve free parameters for shape refinement. The subject of the paper is to develop local positive and monotone irregular surface data interpolation schemes which do not constrain step length and derivatives.

The remaining paper is organized as follows: The cubic Hermite side-vertex interpolation scheme (Nielson, 1979) for irregular surface data is reviewed in Section 2. In Section 3.1,

a rational function with parameters is developed for irregular surface data interpolation. The problems of positive and monotone data interpolation are discussed in Section 3.2 and 3.3 respectively. Section 4 demonstrates the results developed in the previous sections. Section 5 concludes the paper.

2. Cubic Hermite Interpolation of Irregular Surface Data

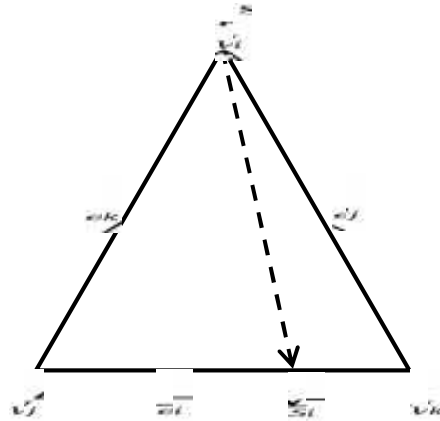


Fig. 1. Location of the vertices and edges of the triangle $\Delta V_i V_j V_k$.

This section provides a review of irregular surface data interpolation scheme proposed by Nielson (1979). The primary requirement was to arrange the data over the triangular grid such that each data point was the vertex of some triangle. The partial derivatives at the vertices of triangle were computed. The data at the three boundaries of a triangle were interpolated by cubic Hermite spline. The rational interpolant of each triangle was a convex combination of radial curves, the curves joining the vertex to the opposite edge of the triangle. The radial curves were also obtained by cubic Hermite interpolation. For any triangle $\Delta V_i V_j V_k$, with barycentric coordinates u , v and w , any point $V = (x, y)$ on the triangle can be expressed as:

$$V = uV_i + vV_j + wV_k, \quad u + v + w = 1 \quad \text{and} \quad u, v, w \geq 0 \quad (1)$$

The Nielson side-vertex interpolant, interpolating the irregular data arranged over a triangular patch was expressed by the following convex combination:

$$P(u, v, w) = \frac{v^2 w^2 RC_i + u^2 w^2 RC_j + u^2 v^2 RC_k}{v^2 w^2 + u^2 w^2 + u^2 v^2}. \quad (2)$$

RC_i , RC_j and RC_k were the radial curves connecting the vertices V_i , V_j and V_k to the opposite edges e_i , e_j and e_k . The order of continuity achieved by Nielson scheme was C^1 .

At the vertices of the triangle $\Delta V_i V_j V_k$ where two of the barycentric coordinates are simultaneously zero, the $P(u, v, w)$ is defined as

$$P(u, 0, 0) = Z_1, \quad P(0, v, 0) = Z_2, \quad P(0, 0, w) = Z_3,$$

where Z_i , $i = 1, 2, 3$ are the ordinate values at the vertices V_i , V_j and V_k respectively.

3. Shape Properties of Irregular Surface Data

In this section, shape preserving interpolation schemes are developed to interpolate the positive and monotone irregular surface data. The data is triangulated by Delaunay triangulation method (Fang and Piegl, 1992). The Nielson side-vertex interpolation scheme by Nielson (1979) is used to interpolate the data over the each triangle with the modification that each boundary curve and each radial curve is the rational cubic function (Sarfraz et al., 1997). Both positivity and monotonicity preserving schemes have developed constraints on parameters to preserve the shape of data.

3.1 Rational interpolant for irregular surface data

Let $\{(x_i, y_i, Z_i), i = 1, 2, 3, \dots, n\}$ be the given set of irregular data arranged over the triangulated grid. Without the loss of generality we shall consider an arbitrary triangle $\Delta V_i V_j V_k$, and will construct the rational interpolant over $\Delta V_i V_j V_k$. The edges of $\Delta V_i V_j V_k$ opposite to the vertices V_i , V_j , V_k are e_i , e_j and e_k respectively. Arbitrary points of the edges e_i , e_j and e_k are denoted by S_i , S_j and S_k . Let RC_i , RC_j and RC_k be the radial curves joining the vertices V_i , V_j , V_k to the edges e_i , e_j and e_k respectively. Z_i , $i = 1, 2, 3$ are the ordinate values at vertices V_i , V_j , V_k respectively. The radial curve RC_i connecting the vertex V_i to the point S_i on the opposite boundary edge e_i is defined as

$$RC_i = \frac{RC_{in}}{RC_{id}}, \quad (3)$$

$$RC_{in} = \{u^3 \alpha_i + (1-u)u^2 A_1\} Z_1 + \{(1-u)^3 \beta_i + (1-u)^2 u A_2\} BC(S_i) + u^2 v \alpha_i D_3 + u^2 w \alpha_i D_4 \\ - u(1-u)v \beta_i D_1 - u(1-u)w \beta_i D_2,$$

$$A_1 = (\alpha_i + \gamma_i), \quad A_2 = (\beta_i + \gamma_i),$$

$$RC_{id} = \alpha_i u^2 + \gamma_i (1-u)u + \beta_i (1-u)^2,$$

$$D_1 = (x_j - x_i) \frac{\partial BC}{\partial x}(S_i) + (y_j - y_i) \frac{\partial BC}{\partial y}(S_i),$$

$$D_2 = (x_k - x_i) \frac{\partial BC}{\partial x}(S_i) + (y_k - y_i) \frac{\partial BC}{\partial y}(S_i),$$

$$D_3 = (x_j - x_i) \frac{\partial F}{\partial x}(V_i) + (y_j - y_i) \frac{\partial F}{\partial y}(V_i),$$

$$D_4 = (x_k - x_i) \frac{\partial F}{\partial x}(V_i) + (y_k - y_i) \frac{\partial F}{\partial y}(V_i),$$

$$BC(S_i) = \frac{s_1^3 \alpha_1 Z_2 + s_1^2 s \{(\alpha_1 + \gamma_1) Z_2 + \alpha_1 d_3\} + s_1 s^2 \{(\beta_1 + \gamma_1) Z_3 - \beta_1 d_4\} + s^3 \beta_1 Z_3}{\alpha_1 s_1^2 + \gamma s_1 s + \beta_1 s^2}, \quad (4)$$

$$d_3 = (x_k - x_j) \frac{\partial F}{\partial x}(V_j) + (y_k - y_j) \frac{\partial F}{\partial y}(V_j),$$

$$d_4 = (x_k - x_j) \frac{\partial F}{\partial x}(V_k) + (y_k - y_j) \frac{\partial F}{\partial y}(V_k),$$

$$s_1 = \frac{v}{v+w}, \quad s = \frac{w}{v+w}.$$

Similarly, the radial curves connecting the vertices V_j and V_k to the edges e_j and e_k are defined as

$$RC_j = \frac{RC_{jn}}{RC_{jd}}, \quad (5)$$

$$RC_{jn} = \{v^3 \alpha_j + (1-v)v^2 B_1\} Z_2 + \{(1-v)^3 \beta_j + (1-v)^2 v B_2\} BC(S_j) + v^2 u \alpha_j D_7 + v^2 w \alpha_j D_8 \\ - v(1-v)u \beta_j D_5 - v(1-v)w \beta_j D_6,$$

$$RC_{jd} = \alpha_j v^2 + 2\gamma_j (1-v)v + \beta_j (1-v)^2,$$

$$B_1 = (\alpha_j + \gamma_j), \quad B_2 = (\beta_j + \gamma_j),$$

$$D_5 = (x_i - x_j) \frac{\partial BC}{\partial x}(S_j) + (y_i - y_j) \frac{\partial BC}{\partial y}(S_j),$$

$$D_6 = (x_k - x_j) \frac{\partial BC}{\partial x}(S_j) + (y_k - y_j) \frac{\partial BC}{\partial y}(S_j),$$

$$D_7 = (x_i - x_j) \frac{\partial F}{\partial x}(V_j) + (y_i - y_j) \frac{\partial F}{\partial y}(V_j),$$

$$D_8 = (x_k - x_j) \frac{\partial F}{\partial x}(V_j) + (y_k - y_j) \frac{\partial F}{\partial y}(V_j),$$

$$BC(S_j) = \frac{r_1^3 \alpha_1 Z_3 + r_1^2 r \{(\alpha_2 + \gamma_2) Z_3 + \alpha_2 d_5\} + r_1 r^2 \{(\beta_2 + \gamma_2) Z_1 - \beta_2 d_6\} + r^3 \beta_2 Z_1}{\alpha_2 r_1^2 + \gamma r_1 r + \beta_1 r^2}, \quad (6)$$

$$d_5 = (x_i - x_k) \frac{\partial F}{\partial x}(V_k) + (y_i - y_k) \frac{\partial F}{\partial y}(V_k),$$

$$d_6 = (x_i - x_k) \frac{\partial F}{\partial x}(V_i) + (y_i - y_k) \frac{\partial F}{\partial y}(V_i),$$

$$r_1 = \frac{w}{u+w}, \quad r = \frac{u}{u+w}.$$

$$RC_k = \frac{RC_{kn}}{RC_{kd}}, \quad (7)$$

$$RC_{kn} = \{w^3 \alpha_k + (1-w)w^2 C_1\} Z_3 + \{(1-w)^3 \beta_k + (1-w)^2 w C_2\} BS(S_k) + w^2 u \alpha_k D_{11} + w^2 v \alpha_k D_{12} \\ - w(1-w)u \beta_k D_9 - w(1-w)v \beta_k D_{10},$$

$$\begin{aligned}
 RC_{kd} &= \alpha_k w^2 + \gamma_k (1-w)w + \beta_k (1-w)^2, \\
 C_1 &= (\alpha_k + \gamma_k), \quad C_2 = (\beta_k + \gamma_k), \\
 D_9 &= (x_i - x_k) \frac{\partial BC}{\partial x}(S_k) + (y_i - y_k) \frac{\partial BC}{\partial y}(S_k), \\
 D_{10} &= (x_j - x_k) \frac{\partial BC}{\partial x}(S_k) + (y_j - y_k) \frac{\partial BC}{\partial y}(S_k), \\
 D_{11} &= (x_i - x_k) \frac{\partial F}{\partial x}(V_k) + (y_i - y_k) \frac{\partial F}{\partial y}(V_k), \\
 D_{12} &= (x_j - x_k) \frac{\partial F}{\partial x}(V_k) + (y_j - y_k) \frac{\partial F}{\partial y}(V_k), \\
 BC(S_k) &= \frac{t_1^3 \alpha_3 Z_1 + t_1^2 t \{(\alpha_3 + \gamma_3) Z_1 + \alpha_3 d_1\} + t_1 t^2 \{(\beta_3 + \gamma_3) Z_2 - \beta_3 d_2\} + t^3 \beta_3 Z_2}{\alpha_3 t_1^2 + \gamma t_1 t + \beta_3 t^2}, \quad (8) \\
 d_1 &= (x_j - x_i) \frac{\partial F}{\partial x}(V_i) + (y_j - y_i) \frac{\partial F}{\partial y}(V_i), \\
 d_2 &= (x_j - x_i) \frac{\partial F}{\partial x}(V_j) + (y_j - y_i) \frac{\partial F}{\partial y}(V_j), \\
 t_1 &= \frac{u}{u+v}, \quad t = \frac{v}{u+v}.
 \end{aligned}$$

The parameters $\alpha_i, \alpha_j, \alpha_k, \alpha_1, \alpha_2, \alpha_3, \beta_i, \beta_j, \beta_k, \beta_1, \beta_2, \beta_3, \gamma_i, \gamma_j, \gamma_k, \gamma_1, \gamma_2$ and γ_3 are parameters known as shape design parameters used to modify the shape of surface as required. The rational interpolant over each triangular patch is constructed by substituting the values RC_i, RC_j and RC_k from (3), (5) and (7) in (2).

3.2 Positivity preserving interpolation of irregular surface data

Let $\{(x_i, y_i, Z_i), i=1,2,3,\dots,n\}$ be the given irregular data, positive over the whole domain ($Z_i > 0, i=1,2,3,\dots,n$).

The rational interpolant (2) will be positive if each of the boundary curve defined in (3), (5) and (7) are positive. The radial curves RC_i, RC_j and RC_k can be rearranged as:

$$RC_i = \frac{RC_{in}}{RC_{id}},$$

where

$$\begin{aligned}
 RC_{in} &= \left\{ (1-u)^3 \beta_i + u(1-u)^2 A_2 \right\} BC(S_i) + \alpha_i Z_1 + v G_1 + w G_2 + v^2 G_3 + vw G_4 + w^2 G_5 + v^3 G_6 + \\
 &\quad v^2 w G_7 + vw^2 G_8 + w^3 G_9, \quad (9)
 \end{aligned}$$

$$RC_{id} = \alpha_i u^2 + \gamma_i (1-u)u + \beta_i (1-u)^2, \quad (10)$$

$$G_1 = -3\alpha_i Z_1 + A_1 Z_1 + \alpha_i D_3,$$

$$G_2 = -3\alpha_i Z_1 + A_1 Z_1 + \alpha_i D_4,$$

$$G_3 = \alpha_i Z_1 - 2A_1 Z_1 - \beta_i D_1 - 2\alpha_i D_3,$$

$$G_4 = 6\alpha_i Z_1 - 4A_1 Z_1 - \beta_i D_1 - \beta_i D_2 - 2\alpha_i D_3 - 2\alpha_i D_4,$$

$$G_5 = 3\alpha_i Z_1 - 2A_1 Z_1 - \beta_i D_2 - 2\alpha_i D_4,$$

$$G_6 = -\alpha_i Z_1 + A_1 Z_1 + \beta_i D_1 + \alpha_i D_3,$$

$$G_7 = -3\alpha_i Z_1 + 3A_1 Z_1 + 2\beta_i D_1 + \beta_i D_2 + 2\alpha_i D_3 + \alpha_i D_4,$$

$$G_8 = -3\alpha_i Z_1 + 3A_1 Z_1 + 2\beta_i D_2 + \beta_i D_1 + 2\alpha_i D_4 + \alpha_i D_3,$$

$$G_9 = -\alpha_i Z_1 + A_1 Z_1 + \beta_i D_2 + \alpha_i D_4,$$

$$RC_j = \frac{RC_{jn}}{RC_{jd}},$$

$$RC_{jn} = \left\{ (1-v)^3 \beta_j + (1-v)^2 v B_2 \right\} BC(S_j) + \alpha_j Z_2 + uL_1 + wL_2 + u^2 L_3 + uwL_4 + w^2 L_5 + u^3 L_6 + u^2 wL_7 + uw^2 L_8 + w^3 L_9, \quad (11)$$

$$RC_{jd} = \alpha_j v^2 + \gamma_j (1-v)v + \beta_j (1-v)^2, \quad (12)$$

$$L_1 = -3\alpha_j Z_2 + B_1 Z_2 + \alpha_j D_7,$$

$$L_2 = -3\alpha_j Z_2 + B_1 Z_2 + \alpha_j D_8,$$

$$L_3 = 3\alpha_j Z_2 - 2B_1 Z_2 - \beta_j D_5 - 2\alpha_j D_7,$$

$$L_4 = 6\alpha_j Z_2 - 4B_1 Z_2 - \beta_j D_5 - \beta_j D_6 - 2\alpha_j D_7 - 2\alpha_j D_8,$$

$$L_5 = 3\alpha_j Z_2 - 2B_1 Z_2 - \beta_j D_6 - 2\alpha_j D_8,$$

$$L_6 = -\alpha_j Z_2 + B_1 Z_2 + \beta_j D_5 + \alpha_j D_7,$$

$$L_7 = -3\alpha_j Z_2 + 3B_1 Z_2 + 2\beta_j D_5 + \beta_j D_6 + 2\alpha_j D_7 + \alpha_j D_8,$$

$$L_8 = -3\alpha_j Z_2 + 3B_1 Z_2 + 2\beta_j D_6 + \beta_j D_5 + 2\alpha_j D_8 + \alpha_j D_7,$$

$$L_9 = -\alpha_j Z_2 + B_1 Z_2 + \beta_j D_6 + \alpha_j D_8,$$

$$RC_k = \frac{RC_{kn}}{RC_{kd}},$$

$$RC_{kn} = \left\{ (1-w)^3 \beta_k + (1-w)^2 w C_2 \right\} BC(S_k) + \alpha_k Z_3 + uM_1 + vM_2 + u^2 M_3 + uvM_4 + v^2 M_5 + u^3 M_6 + u^2 vM_7 + uv^2 M_8 + v^3 M_9, \quad (13)$$

$$RC_{kd} = \alpha_k w^2 + \gamma_k (1-w)w + \beta_k (1-w)^2, \quad (14)$$

$$M_1 = -3\alpha_k Z_3 + C_1 Z_3 + \alpha_k D_{11},$$

$$M_2 = -3\alpha_k Z_3 + C_1 Z_3 + \alpha_k D_{12},$$

$$M_3 = 3\alpha_k Z_3 - 2C_1 Z_3 - \beta_k D_9 - 2\alpha_k D_{11},$$

$$M_4 = 6\alpha_k Z_3 - 4C_1 Z_3 - \beta_k D_9 - \beta_k D_{10} - 2\alpha_k D_{11} - 2\alpha_k D_{12},$$

$$M_5 = 3\alpha_k Z_3 - 2C_1 Z_3 - \beta_k D_{10} - 2\alpha_k D_{12},$$

$$M_6 = -\alpha_k Z_3 + C_1 Z_3 + \beta_k D_9 + \alpha_k D_{11},$$

$$M_7 = -3\alpha_k Z_3 + 3C_1 Z_3 + 2\beta_k D_9 + \beta_k D_{10} + 2\alpha_k D_{11} + \alpha_k D_{12},$$

$$M_8 = -3\alpha_k Z_3 + 3C_1 Z_3 + 2\beta_k D_{10} + \beta_k D_9 + 2\alpha_k D_{12} + \alpha_k D_{11},$$

$$M_9 = -\alpha_k Z_3 + C_1 Z_3 + \beta_k D_{10} + \alpha_k D_{12}.$$

$RC_i > 0$ if $RC_{in} > 0$ and $RC_{id} > 0$.

From (9), $RC_{in} > 0$ if $\alpha_i > 0$, $\beta_i > 0$, $BC(S_i) > 0$ and $G_i > 0$, $i = 1, 2, 3, \dots, 9$.

From (10), $RC_{id} > 0$ if $\alpha_i > 0$, $\beta_i > 0$ and $\gamma_i > 0$.

Using Theorem developed in (Sarfraz, Al-Mulhem and Ashraf, 1997), $BC(S_i) > 0$ if

$$\gamma_1 > \text{Max} \left\{ 0, \frac{-d_3}{Z_2}, \frac{d_4}{Z_3} \right\},$$

$G_i > 0$, $i = 1, 2, 3, \dots, 9$ if

$$\gamma_i > \text{Max} \left\{ 0, \frac{2\alpha_i Z_1 - \alpha_i D_3}{Z_1}, \frac{2\alpha_i Z_1 - \alpha_i D_4}{Z_1}, \frac{-\beta_i D_1 - \alpha_i D_3}{Z_1}, \frac{-\beta_i D_2 - \alpha_i D_4}{Z_1}, \right. \\ \left. \frac{-\beta_i (2D_1 + D_2) - \alpha_i (2D_3 + D_4)}{3Z_1}, \frac{-\beta_i (2D_2 + D_1) - \alpha_i (2D_4 + D_3)}{3Z_1} \right\} \text{ and} \\ \gamma_i < \text{Min} \left\{ 0, \frac{\beta_i D_1 + 2\alpha_i D_3 - \alpha_i Z_1}{-2Z_1}, \frac{\beta_i D_2 + 2\alpha_i D_4 - \alpha_i Z_1}{-2Z_1}, \frac{\beta_i (D_1 + D_2) + 2\alpha_i (D_3 + D_4) - 2\alpha_i Z_1}{-4Z_1} \right\}.$$

Similarly, $RC_j > 0$ if $\gamma_2 > \text{Max} \left\{ 0, -\frac{d_5}{Z_3}, \frac{d_6}{Z_1} \right\}$,

$L_i > 0$, $i = 1, 2, 3, \dots, 9$ if

$$\gamma_j > \text{Max} \left\{ 0, \frac{2\alpha_j Z_2 - \alpha_j D_7}{Z_2}, \frac{2\alpha_j Z_2 - \alpha_j D_8}{Z_2}, \frac{-\beta_j D_5 - \alpha_j D_7}{Z_2}, \frac{-\beta_j D_6 - \alpha_j D_8}{Z_2}, \right. \\ \left. \frac{-\beta_j (2D_5 + D_6) - \alpha_j (2D_7 + D_8)}{3Z_2}, \frac{-\beta_j (2D_6 + D_5) - \alpha_j (2D_8 + D_7)}{3Z_2} \right\} \text{ and} \\ \gamma_j < \text{Min} \left\{ 0, \frac{\beta_j D_5 + 2\alpha_j D_7 - \alpha_j Z_2}{-2Z_2}, \frac{\beta_j D_6 + 2\alpha_j D_8 - \alpha_j Z_2}{-2Z_2}, \frac{\beta_j (D_5 + D_6) + 2\alpha_j (D_7 + D_8) - 2\alpha_j Z_2}{-4Z_2} \right\}$$

Similarly, $RC_k > 0$, if $\gamma_3 > \text{Max} \left\{ 0, -\frac{d_1}{Z_1}, \frac{d_2}{Z_2} \right\}$,

$M_i > 0$, $i = 1, 2, 3, \dots, 9$ if

$$\gamma_k > \text{Max} \left\{ 0, \frac{2\alpha_k Z_3 - \alpha_k D_{11}}{Z_3}, \frac{2\alpha_k Z_3 - \alpha_k D_{12}}{Z_3}, \frac{-\beta_k D_9 - \alpha_k D_{11}}{Z_3}, \frac{-\beta_k D_{10} - \alpha_k D_{12}}{Z_3}, \right. \\ \left. \frac{-\beta_k (2D_9 + D_{10}) - \alpha_k (2D_{11} + D_{12})}{3Z_3}, \frac{-\beta_k (2D_{10} + D_9) - \alpha_k (2D_{12} + D_{11})}{3Z_3} \right\} \text{ and} \\ \gamma_k < \text{Min} \left\{ 0, \frac{\beta_k D_9 + 2\alpha_k D_{11} - \alpha_k Z_3}{-2Z_3}, \frac{\beta_k D_{10} + 2\alpha_k D_{12} - \alpha_k Z_3}{-2Z_3}, \frac{\beta_k (D_9 + D_{10}) + 2\alpha_k (D_{11} + D_{12}) - 2\alpha_k Z_3}{-4Z_3} \right\}$$

The above discussion is summarized as:

Theorem 1. The C^1 triangular patch, defined over the triangular domain in (2), is positive if the following sufficient conditions are satisfied:

$$Q_1 < \gamma_i < Q_2, Q_3 < \gamma_j < Q_4, Q_5 < \gamma_k < Q_6,$$

$$\begin{aligned}
 Q_1 &= \text{Max} \left\{ 0, \frac{2\alpha_i Z_1 - \alpha_i D_3}{Z_1}, \frac{2\alpha_i Z_1 - \alpha_i D_4}{Z_1}, \frac{-\beta_i D_1 - \alpha_i D_3}{Z_1}, \frac{-\beta_i D_2 - \alpha_i D_4}{Z_1}, \right. \\
 &\quad \left. \frac{-\beta_i (2D_1 + D_2) - \alpha_i (2D_3 + D_4)}{3Z_1}, \frac{-\beta_i (2D_2 + D_1) - \alpha_i (2D_4 + D_3)}{3Z_1} \right\}, \\
 Q_2 &= \text{Min} \left\{ 0, \frac{\beta_i D_1 + 2\alpha_i D_3 - \alpha_i Z_1}{-2Z_1}, \frac{\beta_i D_2 + 2\alpha_i D_4 - \alpha_i Z_1}{-2Z_1}, \frac{\beta_i (D_1 + D_2) + 2\alpha_i (D_3 + D_4) - 2\alpha_i Z_1}{-4Z_1} \right\}, \\
 Q_3 &= \text{Max} \left\{ 0, \frac{2\alpha_j Z_2 - \alpha_j D_7}{Z_2}, \frac{2\alpha_j Z_2 - \alpha_j D_8}{Z_2}, \frac{-\beta_j D_5 - \alpha_j D_7}{Z_2}, \frac{-\beta_j D_6 - \alpha_j D_8}{Z_2}, \right. \\
 &\quad \left. \frac{-\beta_j (2D_5 + D_6) - \alpha_j (2D_7 + D_8)}{3Z_2}, \frac{-\beta_j (2D_6 + D_5) - \alpha_j (2D_8 + D_7)}{3Z_2} \right\}, \\
 Q_4 &= \text{Min} \left\{ 0, \frac{\beta_j D_5 + 2\alpha_j D_7 - \alpha_j Z_2}{-2Z_2}, \frac{\beta_j D_6 + 2\alpha_j D_8 - \alpha_j Z_2}{-2Z_2}, \frac{\beta_j (D_5 + D_6) + 2\alpha_j (D_7 + D_8) - 2\alpha_j Z_2}{-4Z_2} \right\}, \\
 Q_5 &= \text{Max} \left\{ 0, \frac{2\alpha_k Z_3 - \alpha_k D_{11}}{Z_3}, \frac{2\alpha_k Z_3 - \alpha_k D_{12}}{Z_3}, \frac{-\beta_k D_9 - \alpha_k D_{11}}{Z_3}, \frac{-\beta_k D_{10} - \alpha_k D_{12}}{Z_3}, \right. \\
 &\quad \left. \frac{-\beta_k (2D_9 + D_{10}) - \alpha_k (2D_{11} + D_{12})}{3Z_3}, \frac{-\beta_k (2D_{10} + D_9) - \alpha_k (2D_{12} + D_{11})}{3Z_3} \right\}, \\
 Q_6 &= \text{Min} \left\{ 0, \frac{\beta_k D_9 + 2\alpha_k D_{11} - \alpha_k Z_3}{-2Z_3}, \frac{\beta_k D_{10} + 2\alpha_k D_{12} - \alpha_k Z_3}{-2Z_3}, \frac{\beta_k (D_9 + D_{10}) + 2\alpha_k (D_{11} + D_{12}) - 2\alpha_k Z_3}{-4Z_3} \right\}.
 \end{aligned}$$

3.3 Monotonicity preserving interpolation of irregular surface data

In this section, we shall establish the sufficient conditions for the monotonicity of rational interpolant (2) in an arbitrary direction $d = \lambda_1 V_1 + \lambda_2 V_2 + \lambda_3 V_3$, $\lambda_1 + \lambda_2 + \lambda_3 = 0$.

(Floater and Peña, 2000) stated:

Definition 1. A function $f(x, y)$ is said to be strictly monotone in any direction d if

$$D_d f(x, y) > 0,$$

where D_d denotes the directional derivative along the direction d . \square

Let $\{(x_i, y_i, Z_i), i=1,2,3\}$ be the monotone irregular surface data defined over the triangle $\Delta V_i V_j V_k$ obeying the restrictions that if $x_i < x_j$ and $y_i < y_j$ then $Z_i < Z_j$ and $Z_l^x > 0$ and $Z_l^y > 0$, $l = i, j, k$.

The directional derivative of (2) along the direction $d = \lambda_1 V_1 + \lambda_2 V_2 + \lambda_3 V_3$, with $\lambda_1 + \lambda_2 + \lambda_3 = 0$ is

$$D_d P = \lambda_1 \frac{\partial P}{\partial u} + \lambda_2 \frac{\partial P}{\partial v} + \lambda_3 \frac{\partial P}{\partial w} = \frac{(D_d P)_N}{(D_d P)_D}, \quad (15)$$

$$(D_d P)_N = u^2 v^4 w^2 E_1 + u^4 v^2 w^2 E_2 + u^4 v^4 E_3 + v^4 w^4 E_4 + uv^2 w^4 E_5 + u^2 v^2 w^4 E_6 + uv^4 w^2 E_7 + u^4 w^4 E_8 + u^2 v w^4 E_9 + u^2 v^4 w E_{10} + u^4 v w^2 E_{11} + u^4 v^2 w E_{12}, \quad (16)$$

$$(D_d P)_D = (u^2 v^2 + v^2 w^2 + w^2 u^2)^2 \quad (17)$$

E_i , $i = 1, 2, 3, \dots, 12$ can be obtained by simple computation involved in (15). Using the Definition (1), the rational interpolant (2) is monotone if $D_d P > 0$.

$(D_d P)_D > 0$ is positive always, whereas, $(D_d P)_N > 0$ if $E_i > 0$, $i = 1, 2, 3, \dots, 12$.

$E_i > 0$, $i = 1, 2, 3, \dots, 12$ if

$RC_{id} > 0$, with $i = 1, 2, 3$, $\eta_i > 0$, $\chi_i > 0$, $\delta_i > 0$; $i = 1, 2, 3, \dots, 17$.

$\lambda_1(RC_j - RC_i) > 0$, $\lambda_1(RC_k - RC_i) > 0$, $\lambda_2(RC_k - RC_j) > 0$, $\lambda_3(RC_i - RC_k) > 0$,

$\lambda_2(RC_k - RC_j) > 0$, $\lambda_3(RC_j - RC_k) > 0$, $\eta_i > 0$, $\chi_i > 0$, $\delta_i > 0$; $i = 1, 2, 3, \dots, 17$.

$\gamma_1 > \text{Max}\{0, \text{Con}_i, 1 \leq i \leq 8\}$, $\gamma_2 > \text{Max}\{0, \text{Con}_i, 9 \leq i \leq 16\}$,

$\gamma_3 > \text{Max}\{0, \text{Con}_i, 17 \leq i \leq 24\}$, $\gamma_4 > \text{Max}\{0, \text{Con}_i, 25 \leq i \leq 30\}$,

$\gamma_5 > \text{Max}\{0, \text{Con}_i, 31 \leq i \leq 36\}$, $\gamma_6 > \text{Max}\{0, \text{Con}_i, 37 \leq i \leq 42\}$,

where

$$\text{Con}_1 = \frac{\beta_i(\lambda_2 D_1 + \lambda_3 D_2) - 2\lambda_1 \beta_i(Z_1 - BC(S_i))}{2\lambda_1(Z_1 - BC(S_i)) + (\lambda_2 D_3 + \lambda_3 D_4)},$$

$$\text{Con}_2 = \frac{-4\lambda_1 \alpha_i(Z_1 - BC(S_i)) - (\lambda_2 D_3 + \lambda_3 D_4)\alpha_i}{(Z_1 - BC(S_i))\lambda_1},$$

$$\text{Con}_3 = \frac{(\lambda_2 D_1 + \lambda_3 D_2)\beta_i - \lambda_1 \alpha_i(Z_1 - BC(S_i))}{\lambda_1(Z_1 - BC(S_i))},$$

$$\text{Con}_4 = \frac{\beta_i(\lambda_2 D_1 + \lambda_3 D_2) - 2\lambda_1 \alpha_i(Z_1 - BC(S_i))}{2\lambda_1(Z_1 - BC(S_i))},$$

$$\text{Con}_5 = -\frac{\beta_i D_1}{D_3},$$

$$\text{Con}_6 = -\frac{\beta_i D_2}{D_4},$$

$$\text{Con}_7 = \frac{-\beta_i(D_1 + 2D_3)}{D_3},$$

$$\text{Con}_8 = \frac{-\beta_i(D_2 + 2D_4)}{D_4},$$

$$\text{Con}_9 = \frac{\beta_j(\lambda_4 D_5 + \lambda_3 D_6) - 2\lambda_2 \beta_j(Z_2 - BC(S_j))}{2\lambda_2(Z_2 - BC(S_j)) + (\lambda_4 D_7 + \lambda_3 D_8)},$$

$$\text{Con}_{10} = \frac{-4\lambda_2 \alpha_j(Z_2 - BC(S_j)) - (\lambda_4 D_7 + \lambda_3 D_8)\alpha_j}{(Z_2 - BC(S_j))\lambda_2},$$

$$\begin{aligned}
 Con_{11} &= \frac{(\lambda_1 D_5 + \lambda_3 D_6) \beta_j - \lambda_2 \alpha_j (Z_2 - BC(S_j))}{\lambda_2 (Z_2 - BC(S_j))}, \\
 Con_{12} &= \frac{\beta_j (\lambda_1 D_5 + \lambda_3 D_6) - 2 \lambda_2 \alpha_j (Z_2 - BC(S_j))}{2 \lambda_2 (Z_2 - BC(S_j))}, \\
 Con_{13} &= -\frac{\beta_j D_5}{D_7}, \\
 Con_{14} &= -\frac{\beta_j D_6}{D_8}, \\
 Con_{15} &= \frac{-\beta_j (D_5 + 2D_7)}{D_7}, \\
 Con_{16} &= \frac{-\beta_j (D_6 + 2D_8)}{D_8}, \\
 Con_{17} &= \frac{\beta_k (\lambda_1 D_9 + \lambda_2 D_{10}) - 2 \lambda_3 \beta_k (Z_3 - BC(S_k))}{2 \lambda_3 (Z_3 - BC(S_k)) + (\lambda_1 D_{11} + \lambda_2 D_{12})}, \\
 Con_{18} &= \frac{-4 \lambda_3 \alpha_k (Z_3 - BC(S_k)) - (\lambda_1 D_{11} + \lambda_2 D_{12}) \alpha_k}{(Z_3 - BC(S_k)) \lambda_3}, \\
 Con_{19} &= \frac{(\lambda_1 D_9 + \lambda_2 D_{10}) \beta_k - \lambda_3 \alpha_k (Z_3 - BC(S_k))}{\lambda_3 (Z_3 - BC(S_k))}, \\
 Con_{20} &= \frac{\beta_k (\lambda_1 D_9 + \lambda_2 D_{10}) - 2 \lambda_3 \alpha_k (Z_3 - BC(S_k))}{2 \lambda_3 (Z_3 - BC(S_k))}, \\
 Con_{21} &= -\frac{\beta_k D_9}{D_{11}}, \\
 Con_{22} &= -\frac{\beta_k D_{10}}{D_{12}}, \\
 Con_{23} &= \frac{-\beta_k (D_9 + 2D_{11})}{D_{11}}, \\
 Con_{24} &= \frac{-\beta_k (D_{10} + 2D_{12})}{D_{12}}, \\
 Con_{25} &= \frac{\lambda_2 \alpha_1 d_3 - 2 \lambda_3 \beta_1 (Z_3 - Z_2) + 2 \lambda_3 \beta_1 d_4}{2 \lambda_3 (Z_3 - Z_2)}, \\
 Con_{26} &= \frac{\lambda_3 \alpha_1 d_3 + (2 \lambda_2 - \lambda_3) \alpha_1 (Z_3 - Z_2)}{\lambda_3 (Z_3 - Z_2)}, \\
 Con_{27} &= \frac{\lambda_3 \alpha_1 d_4 - 3 \lambda_3 \alpha_1 (Z_3 - Z_2) - 2 \lambda_2 \alpha_1 d_4}{-\lambda_3 d_4}, \\
 Con_{28} &= \frac{-\lambda_2 \beta_1 d_4 + 2 \beta_1 \lambda_3 (Z_2 - Z_3) - \lambda_2 \beta_1 (Z_2 - Z_3)}{\lambda_2 (Z_2 - Z_3)},
 \end{aligned}$$

$$\begin{aligned}
 Con_{29} &= \frac{2\lambda_3\beta_1d_3 - \lambda_2\beta_1d_3 - 3\lambda_2\beta_1(Z_2 - Z_3)}{\lambda_2d_3}, \\
 Con_{30} &= \frac{-2\lambda_2\alpha_1d_3 - \lambda_3\beta_1d_4 + 2\lambda_2\alpha_1(Z_3 - Z_2)}{2\lambda_2(Z_2 - Z_3)}, \\
 Con_{31} &= \frac{\lambda_3\alpha_2d_5 - 2\lambda_1\beta_2(Z_1 - Z_3) + 2\lambda_1\beta_2d_6}{2\lambda_1(Z_1 - Z_3)}, \\
 Con_{32} &= \frac{\lambda_1\alpha_2d_5 + (2\lambda_3 - \lambda_1)\alpha_2(Z_1 - Z_3)}{\lambda_1(Z_1 - Z_3)}, \\
 Con_{33} &= \frac{\lambda_1\alpha_2d_6 - 3\lambda_1\alpha_2(Z_1 - Z_3) - 2\lambda_3\alpha_2d_6}{-\lambda_1d_6}, \\
 Con_{34} &= \frac{-\lambda_3\beta_2d_6 + 2\beta_2\lambda_1(Z_3 - Z_1) - \lambda_3\beta_2(Z_3 - Z_1)}{\lambda_3(Z_3 - Z_1)}, \\
 Con_{35} &= \frac{2\lambda_1\beta_2d_5 - \lambda_3\beta_2d_5 - 3\lambda_3\beta_2(Z_3 - Z_1)}{\lambda_3d_5}, \\
 Con_{36} &= \frac{-2\lambda_3\alpha_2d_5 - \lambda_1\beta_2d_6 + 2\lambda_3\alpha_2(Z_1 - Z_3)}{2\lambda_3(Z_3 - Z_1)}, \\
 Con_{37} &= \frac{\lambda_1\alpha_3d_1 - 2\lambda_2\beta_3(Z_2 - Z_1) + 2\lambda_2\beta_3d_2}{2\lambda_2(Z_2 - Z_1)}, \\
 Con_{38} &= \frac{\lambda_2\alpha_3d_1 + (2\lambda_1 - \lambda_2)\alpha_3(Z_1 - Z_2)}{\lambda_2(Z_2 - Z_1)}, \\
 Con_{39} &= \frac{\lambda_2\alpha_3d_2 - 3\lambda_2\alpha_3(Z_2 - Z_1) - 2\lambda_1\alpha_3d_2}{-\lambda_2d_2}, \\
 Con_{40} &= \frac{-\lambda_1\beta_3d_2 + 2\beta_3\lambda_2(Z_1 - Z_2) - \lambda_1\beta_3(Z_1 - Z_2)}{\lambda_1(Z_1 - Z_2)}, \\
 Con_{41} &= \frac{2\lambda_2\beta_3d_1 - \lambda_1\beta_3d_1 - 3\lambda_1\beta_3(Z_1 - Z_2)}{\lambda_1d_1}, \\
 Con_{42} &= \frac{-2\lambda_1\alpha_3d_1 - \lambda_2\beta_3d_2 + 2\lambda_1\alpha_3(Z_2 - Z_1)}{2\lambda_1(Z_1 - Z_2)}.
 \end{aligned}$$

The above discussion is summarized as:

Theorem 2. The triangular patch P , defined over the triangular domain, in (2), is monotone in the direction $d = \lambda_1V_1 + \lambda_2V_2 + \lambda_3V_3$, with $\lambda_1 + \lambda_2 + \lambda_3 = 0$ if the following conditions are satisfied

$$\begin{aligned}
 \gamma_1 &= p_1 + \text{Max}\{0, Con_i, 1 \leq i \leq 8\}, & \gamma_2 &= p_2 + \text{Max}\{0, Con_i, 9 \leq i \leq 16\}, \\
 \gamma_3 &= p_3 + \text{Max}\{0, Con_i, 11 \leq i \leq 24\}, & \gamma_4 &= p_4 + \text{Max}\{0, Con_i, 25 \leq i \leq 30\}, \\
 \gamma_5 &= p_5 + \text{Max}\{0, Con_i, 31 \leq i \leq 36\}, & \gamma_6 &= p_6 + \text{Max}\{0, Con_i, 37 \leq i \leq 42\},
 \end{aligned}$$

where $p_i > 0$, $i = 1, 2, 3, \dots, 6$.

4. Numerical Examples

In this section, the shape preserving interpolating scheme developed in Section 3 is implemented on some functions.

Example 1. The positive irregular surface data is generated from the following positive function

$$F_1(x, y) = (xy - 1)^2 + 0.2,$$

and the domain is restricted to $[-4, 4] \times [-4, 4]$. Figs. 2 and 3 show the Delaunay triangulation of the domain and linear interpolation of the positive irregular surface data generated from the function $F_1(x, y)$. Fig.4 is generated from the interpolating scheme, discussed in Section 3 for the value of shape parameters $\alpha_1 = \alpha_2 = \alpha_3 = \beta_1 = \beta_2 = \beta_3 = \alpha_i = \alpha_j = \alpha_k = \beta_i = \beta_j = \beta_k = 1$ and $\gamma_i = \gamma_j = \gamma_k = \gamma_1 = \gamma_2 = \gamma_3 = 2$ which reduces the given interpolant to the cubic Hermite triangular interpolant. Fig. 5 is another view of Fig.4. Figs. 4 and 5 show that some part of the surface lie below the plane $Z = 0$, it means that it only interpolates the data but does not preserve the shape of the positive irregular surface data. Fig.6. is generated from Theorem 1 with free parameters $\alpha_1 = \alpha_2 = \alpha_3 = \beta_1 = \beta_2 = \beta_3 = \alpha_i = \alpha_j = \alpha_k = \beta_i = \beta_j = \beta_k = 1.3$. In Fig.6, it is to be noted that the shape of the positive irregular surface data is preserved. Fig.7 provides another view of Fig.6.

Example 2. The next example is also for the positive irregular surface data generated from the following function

$$F_2(x, y) = x^2y + e^{xy}\sin(y),$$

and domain is restricted to $[0, 3] \times [0, 3]$. Figs. 8 and 9 represent the Delaunay triangulation and linear interpolation of the irregular surface data generated from the function $F_2(x, y)$. Fig.10 is the cubic Hermite triangular surface of the given function and we observed that cubic Hermite did not preserve the shape of positive data. Fig.11 is another view of Fig.10. Fig.12 is generated from the Theorem 1 for the free parameters are $\alpha_1 = \alpha_2 = \alpha_3 = \beta_1 = \beta_2 = \beta_3 = \alpha_i = \alpha_j = \alpha_k = \beta_i = \beta_j = \beta_k = 1.22$ and the surface generated from this scheme is positive. Fig.13 is another view of Fig.12.

Example 3. The monotone irregular surface data are generated from the following function

$$F_3(x, y) = x + \ln(x^2 + y^2) - 1.5,$$

and domain is restricted to $[1, 10] \times [1, 10]$. Figs. 14 and 15 are the Delaunay triangulation and linear interpolation of the monotone irregular surface data. Fig.16. is the cubic Hermite triangular surface and by using the Hermite interpolant, the surface deviates from its monotone behaviour. Figs. 17 and 18 are the xz -view and yz -view of the Fig.16. Fig.19. is generated from Theorem 2 for the values of free parameters $\alpha_1 = \alpha_2 = \alpha_3 = \beta_1 = \beta_2 = \beta_3 = \alpha_i = \alpha_j = \alpha_k = \beta_i = \beta_j = \beta_k = 0.001$. This scheme has preserved the monotone shape of irregular surface data. Figs. 20 and 21 are the xz -view and yz -view of Fig.19.

Example 4. The fourth example is also for the monotone irregular surface data generated from the following function

$$F_4(x, y) = \ln(2x + y + 1).$$

The domain is restricted to $[0, 5] \times [0, 5]$. Fig. 22 and 23 are the Delaunay triangulation of the domain and linear interpolation of the irregular data generated from the function $F_4(x, y)$. Fig.24. is the cubic Hermite triangular surface and it does not preserve the shape of monotone data. Figs 25 and 26 are the xz -view and yz -view of the Fig.24. Fig.27 is generated from Theorem 2 and values of shape parameters are $\alpha_1 = \alpha_2 = \alpha_3 = \beta_1 = \beta_2 = \beta_3 = \alpha_i = \alpha_j = \alpha_k = \beta_i = \beta_j = \beta_k = 0.05$. Monotone shape of the data has been preserved in Fig.27. Figs. 28 and 29 are the xz -view and yz -view of Fig.27.

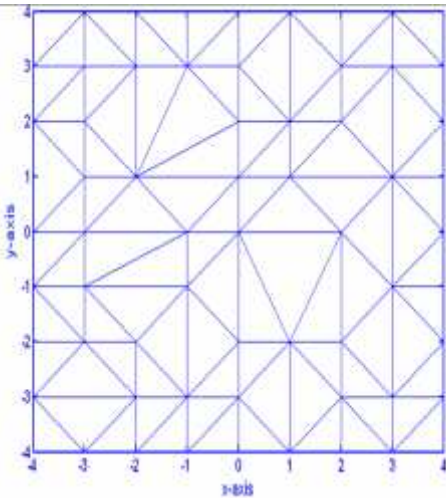


Fig.2. Triangulation of the domain $F_1(x, y)$.

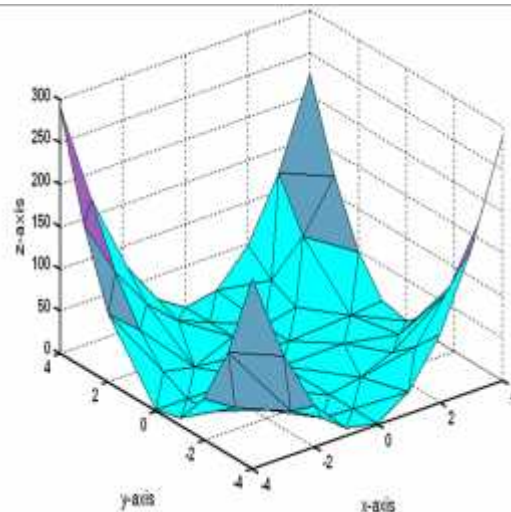


Fig.3. Linear interpolation of $F_1(x, y)$.

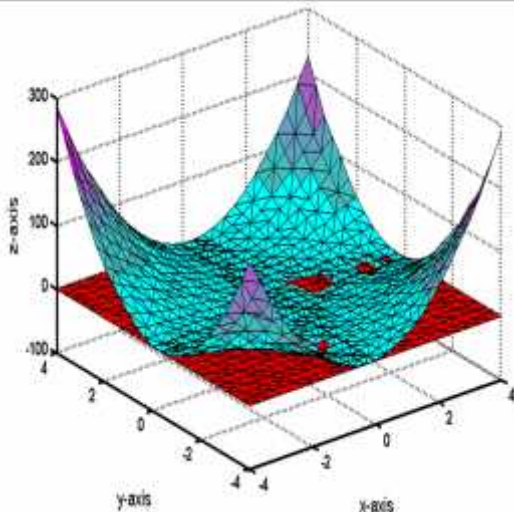


Fig.4. Hermite triangular surface of $F_1(x, y)$.

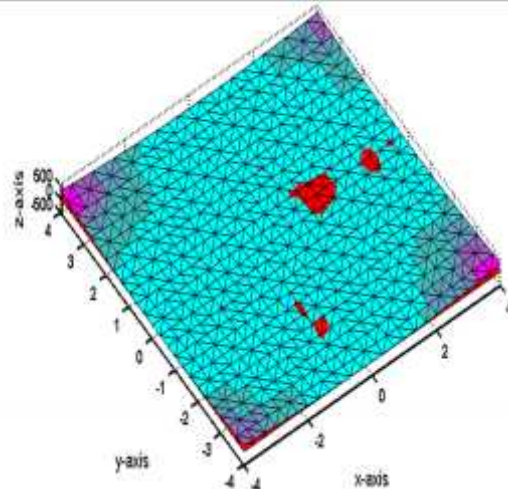


Fig.5. Another view of Fig.4.

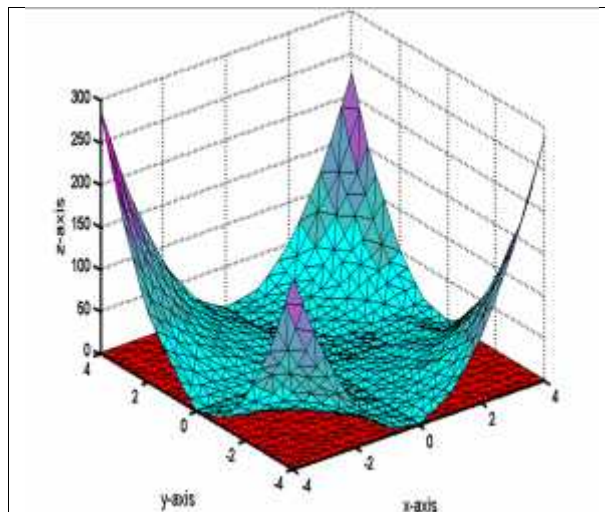


Fig.6. Positive side-vertex interpolant of $F_1(x, y)$.

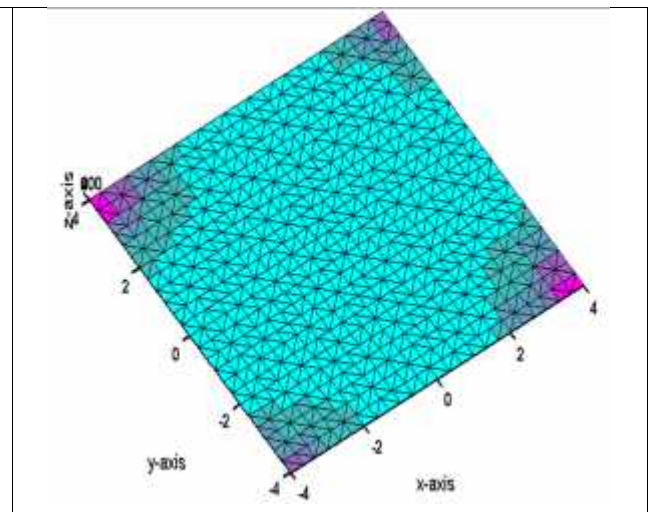


Fig.7. Another view of Fig.6.

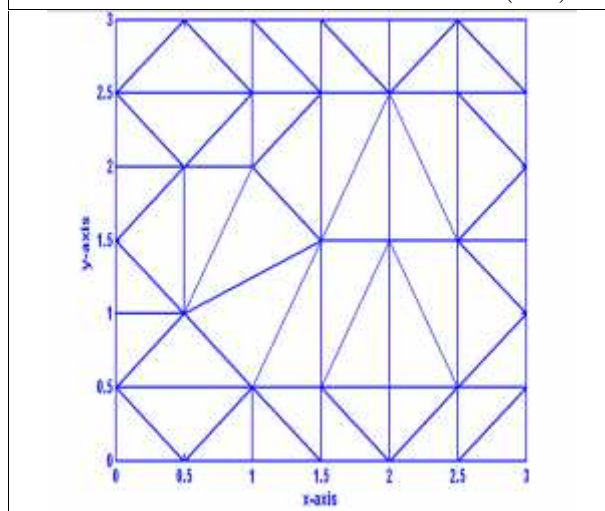


Fig.8. Triangulation of the domain $F_2(x, y)$.

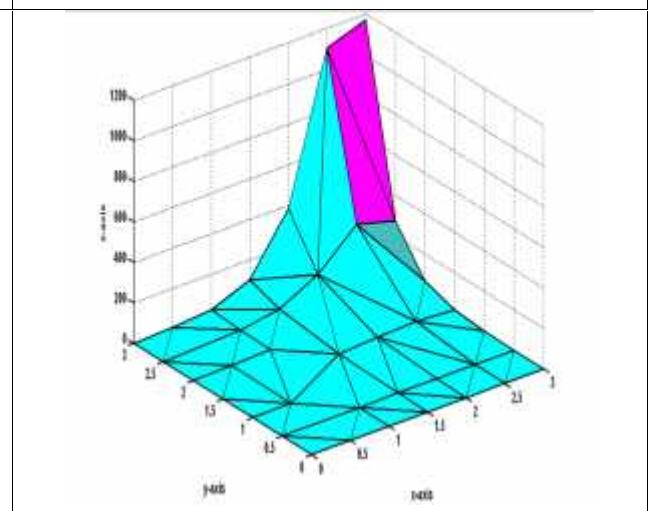


Fig.9. Linear interpolation of $F_2(x, y)$.

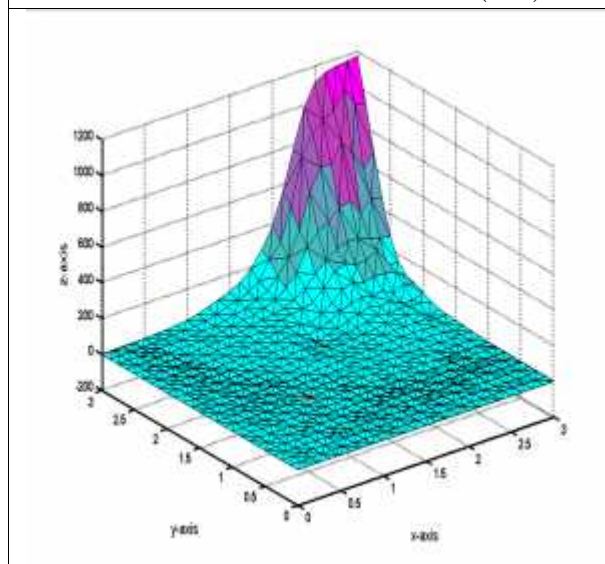


Fig.10. Hermite triangular surface of $F_2(x, y)$.

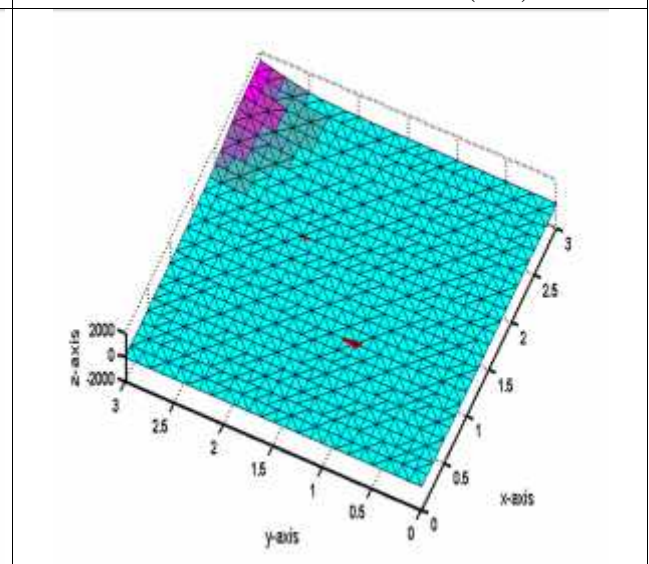


Fig.11. Another view of Fig.10.

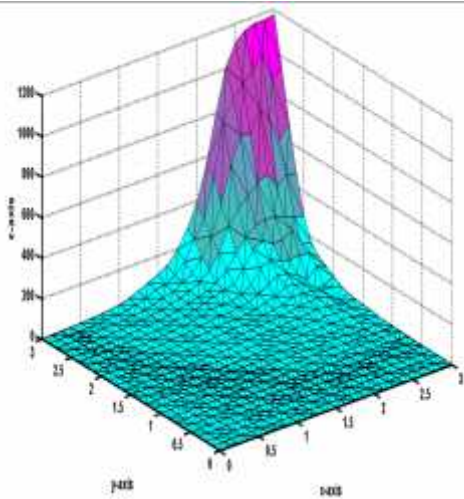


Fig.12. Positive side-vertex interpolant of $F_2(x, y)$.

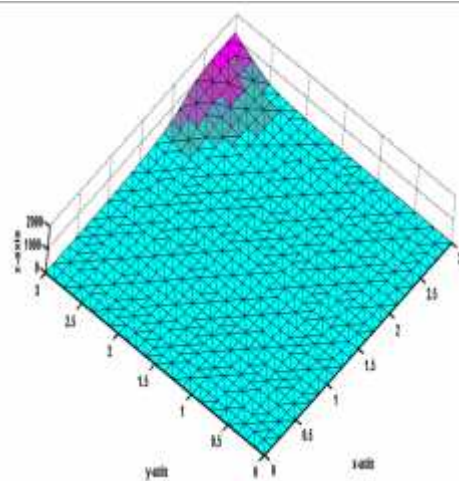


Fig.13. Another view of Fig.12.

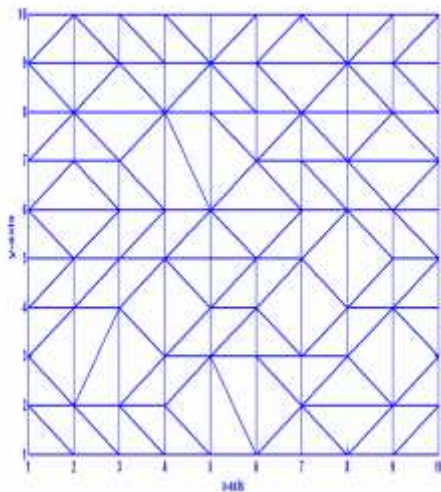


Fig.14. Triangulation of the domain $F_3(x, y)$.

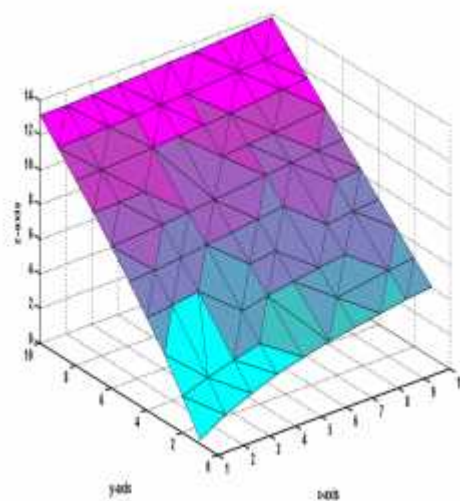


Fig.15. Linear interpolation of $F_3(x, y)$.

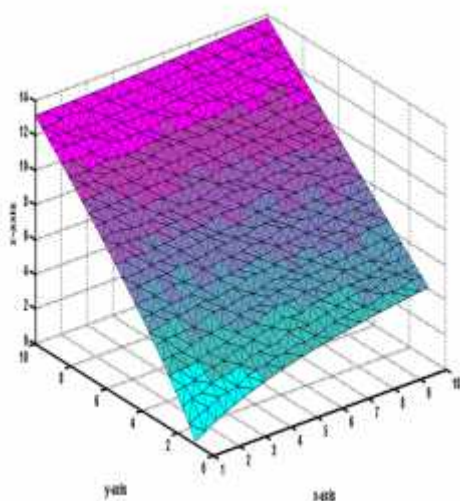


Fig.16. Hermite triangular surface of $F_3(x, y)$.

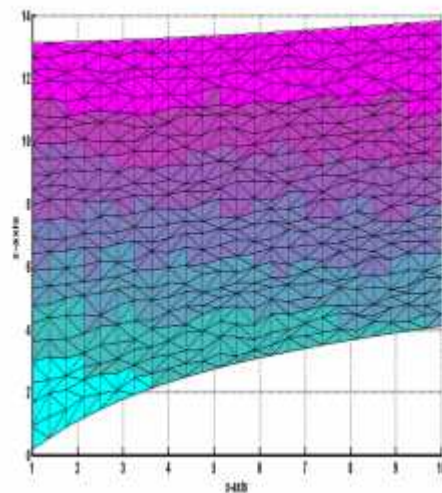


Fig.17. xz-view of Fig.16.

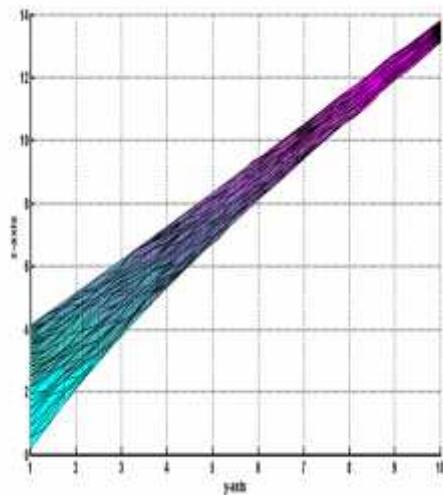


Fig.18. yz-view of Fig.16.

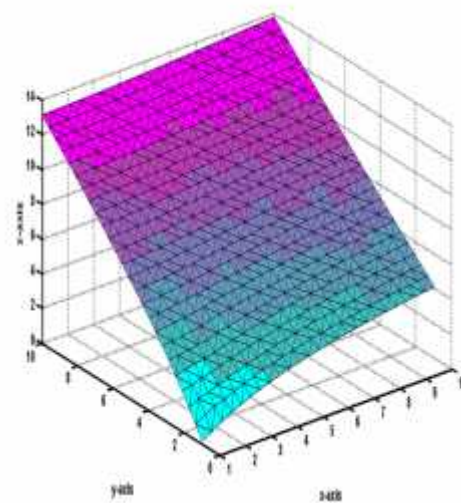


Fig.19. Monotone side-vertex interpolant of $F_3(x, y)$.

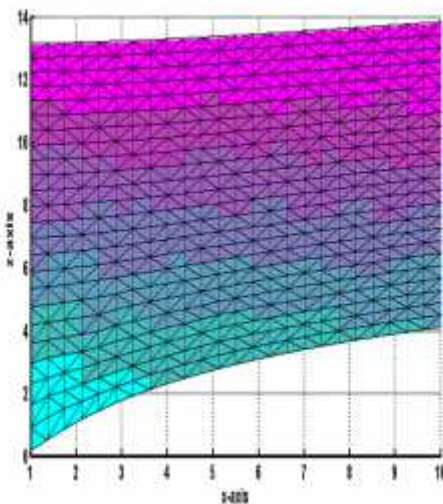


Fig.20. xz-view of Fig.19.

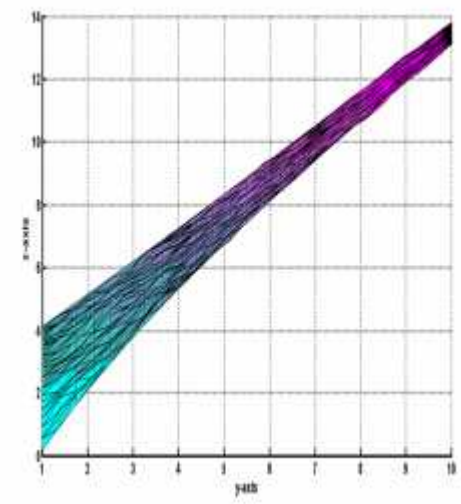


Fig.21. yz-view of Fig.19.

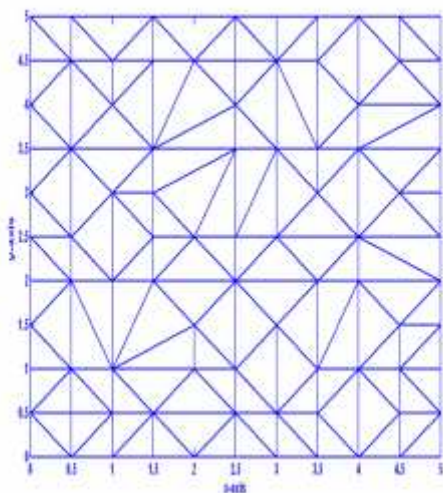


Fig.22. Triangulation of the domain of $F_4(x, y)$.

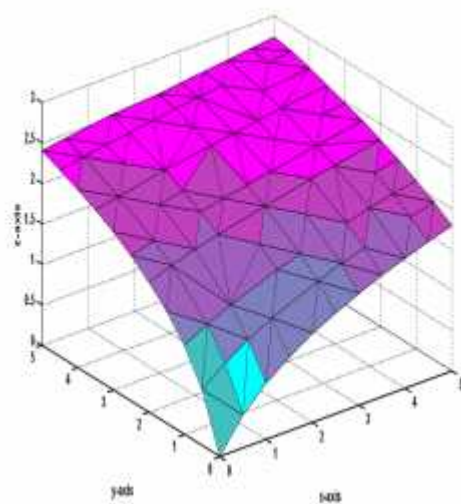


Fig.23. Linear interpolation of $F_4(x, y)$.

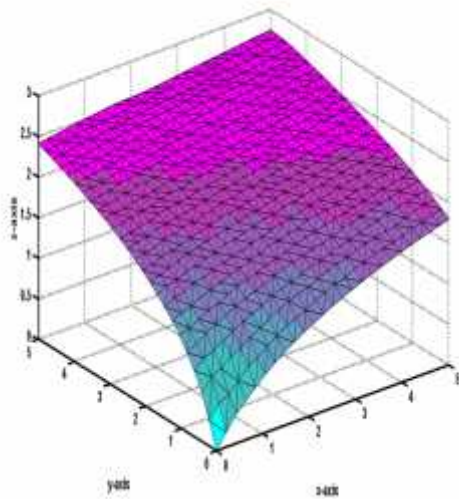


Fig.24. Hermite triangular surface of $F_4(x, y)$

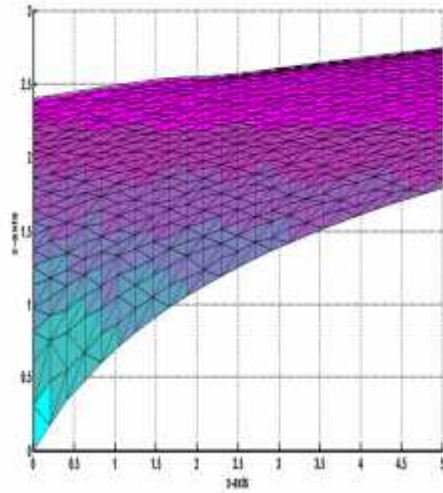


Fig.25. xz-view of Fig.24.

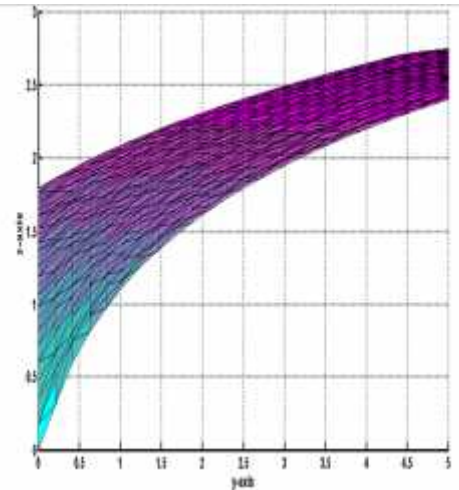


Fig.26. yz-view of Fig.24.

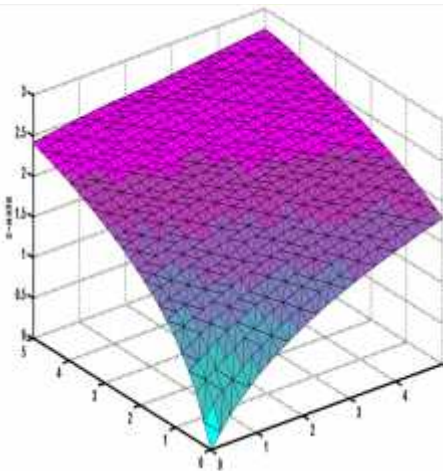


Fig.27. Monotone side-vertex interpolant of $F_4(x, y)$.

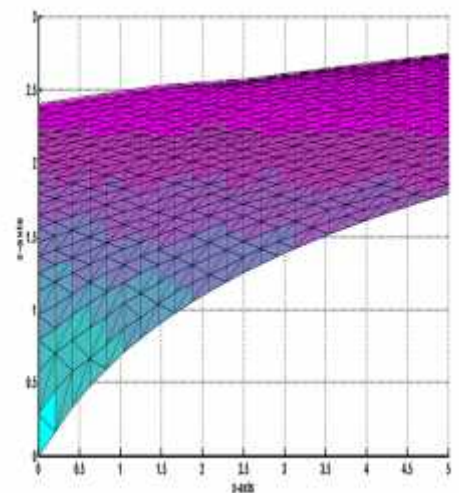


Fig.28. xz-view of Fig.27.

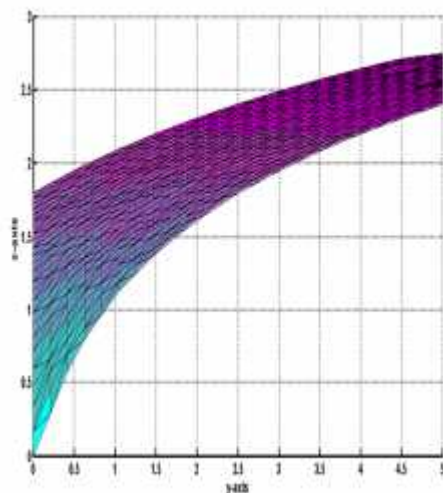


Fig.29. yz-view of Fig.27.

5. Conclusion

In (Beatson and Ziegler, 1985, Chan and Ong, 2001, Piah, Goodman and Unsworth, 2005, Piah, Saaban and Majid, 2006), constraints are derived on derivatives. When derivatives are given with data then these schemes are not helpful. But in this paper, the scheme is acceptable to both data with and without derivatives. In (Hussain, et al, 2009), a piecewise rational cubic function with one free parameter is used. This scheme does not give the freedom to the user to modify the shape of the data. In this paper, 12 shape parameters in each triangular patch are free for user choice to modify the shape of the data.

The monotonicity preserving scheme presented in (Beliakov, 2005) is only applicable to Lipschitz continuous functions. In (Han and Schumaker, 1997) the given irregular data is arranged to rectangular grid but the schemes of this paper are acceptable to both rectangular and triangular grid.

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