

## ***Solve Inverse problem of Fractal Rendering 3D Shapes.***

Mushtaq Kareem AbdalRahem  
Karbala University  
MSc. in Mathematics and Computer Applications

### **Abstract**

The inverse problem of fractal shapes can be considered as the main difficulty in fractal geometry, because the problem of determining and identifying the parameters of the affine mappings that constitute the Iterated Function System (IFS).

The purpose of this paper is to formulate and solve the inverse problem of Fractal Rendering 3D Shapes using a new approach in optimization theory. This can be made by finding and generating parameters of a set of affine maps, contractive mapping; which is IFS for patches; that repeatedly iterative numerically by using the random iteration algorithm

### **المستخلص**

المسألة العكسية للأشكال الكسورية يمكن أن توصف بالصعوبة الرئيسية في الهندسة الكسورية، بسبب وجود مشكلة في حساب وتمييز معاملات مخططات التقريب الذي يشكل نظام الوظيفة المتكرر (IFS). إن غرض هذا البحث هو صياغة وحل المسألة العكسية للأشكال الكسورية المعادة الثلاثية الأبعاد باستخدام طريقة جديدة في نظرية الأمثلية. ويمكن هذا بإيجاد وتوليد الثوابت أو العوامل لمجموعة الدوال، الذي يشكل نظام الوظيفة المتكرر (IFS) والذي يتكرر عددياً باستخدام خوارزمية التكرار العشوائية.

### **1- Introduction**

Fractals discovered by Benoit Mandelbrot [4] in 1970s have changed the way we see everything. Natural objects such as clouds, plants, landscapes and many other objects, complex in shape, can be efficiently modeled using this new tool. The main property of fractals – self-similarity is the crucial feature used in finding fractal parameters needed in modeling process.

Fractals possess many different aspects through which they can be characterized. Among these are:

- Self-similarity.
- Non-integer dimensionality.
- Being attractors of some peculiar dynamical system.

Self-similarity means that the whole object is similar to its parts. For commonly known fractals such as Koch's curve, Cantor's set and Sierpinski's gasket self-similarity property is evidently seen. But smooth curves or patches usually we do not treat as fractals although in reality they are, as it has been shown in [3],[11]. Eg. Self – similarity of Bézier curves or patches is not obvious. Based on self-similarity property one can define IFS (Iterated Function System) that can be further used to generate iteratively the so-called attractor (fixed point of IFS) that is a fractal. So fractals may be considered as attractors of IFSs. It is worth mentioning that if one knows IFS then one can generate its attractor easily using one of the methods – deterministic or probabilistic one. But the solution of the inverse problem to find IFS of any attractor, in general, is not solved yet.

Inspired by [3], [10] in this paper, using self-similarity property we derive formulas that give fractal description of patches based on 3 control points.

Having IFSs for fractal rendering of patches one may use it for fractal generation of any 3D shape that may be presented as a finite collection of single patches. Precisely speaking, we use the

so-called PIFS (Partitioned IFS), not IFS, because every patch is modeled separately by its own IFS. We present examples of fractal rendering both for single patches and for some more complex 3D shapes consisting of several patches.

**2- The Inverse Problem**

The so – called "*Inverse Problem*" of fractal shapes can be considered as the main difficulty in fractal geometry, because the problem of determining and identifying the parameters of the affine mappings that constitute the Iterated Function System (IFS). Originally, [2] solved it by laborious human manipulation (requiring about 100 hours of effort per image). Early automated attempts (e.g. [4]) relied upon matching power moments of candidate IFS with those of the target image. They require huge (and slow) searches through parameter spaces, and tend to work poorly for target images of dimension 2 and higher. Later attempts (e.g. [1], [8]) simplified the problem drastically by making assumptions about the character of the transformations.

**3- Fractal Rendering of 3D Shapes**

Fractal rendering is progressive in nature, in contrary to geometrical modeling based on control points patches. Progressiveness is a very desirable property during transmitting of graphical information through the net. So

then, fractal rendering may be of the interests for practice. The results presented in the paper may be treated as an attempt of extension of fractal rendering method from 2D to 3D. Unfortunately, the results presented here are not so complete as for 2D case for which a fractal rendering pipeline [13] has been proposed. Namely, any 2D contour could be there automatically splatted into a finite number of segments with known IFS for each one. Exemplary two shapes (a rabbit and a vase) presented in this paper have been modeled manually. Additionally, one more complex shape, the famous teapot from Utah, also had been rendered successfully in the fractal way. Some examples of fractal modeling of 2D shapes are presented in previous authors' papers [13], while 3D fractal shapes are displayed on [7].

**4- IFS For Patches**

Below in this section we give formulas that define IFSs for fractal rendering of 3-points-based patches. They seem to form a collection of patches sufficient to build up a wide class of real and artificial 3D shapes. Also example of patches of different kinds rendered using deterministic or probabilistic methods are presented [13].

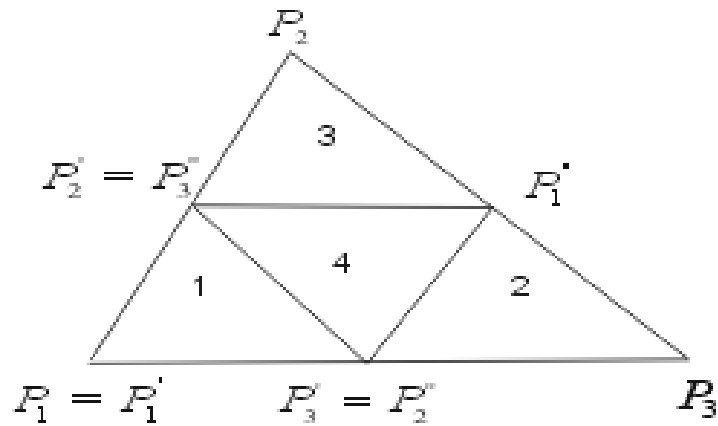
Any triangle in  $R^3$  with vertices  $P_1=[x_1,y_1,z_1]$ ,  $P_2=[x_2,y_2,z_2]$ ,  $P_3=[x_3,y_3,z_3]$  can be divided into four smaller ones. Applying the mid-point subdivision strategy we can define the following four subdivision matrices:

$$\begin{aligned}
 B_1 &= \begin{bmatrix} 1 & 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}, & B_2 &= \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix} \\
 B_3 &= \begin{bmatrix} 1 & 1 & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}, & B_4 &= \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}
 \end{aligned}$$

Non-singular quadratic matrix  $P$  has the form:

$$P = \begin{bmatrix} x_1 & y_1 & z_1 & 1 \\ x_2 & y_2 & z_2 & 1 \\ x_3 & y_3 & z_3 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix} \text{ x y z}$$

It should be pointed out that vertices of small triangles are enumerated in the same way (in clockwise direction) as in the input triangle, as shown in Fig.1. Similarity between the input triangle and the 4<sup>th</sup> small triangle is defined by  $B_4P$ . The same is with the left three small triangles.



*Fig.1. Subdivision of a triangle into four smaller ones.*

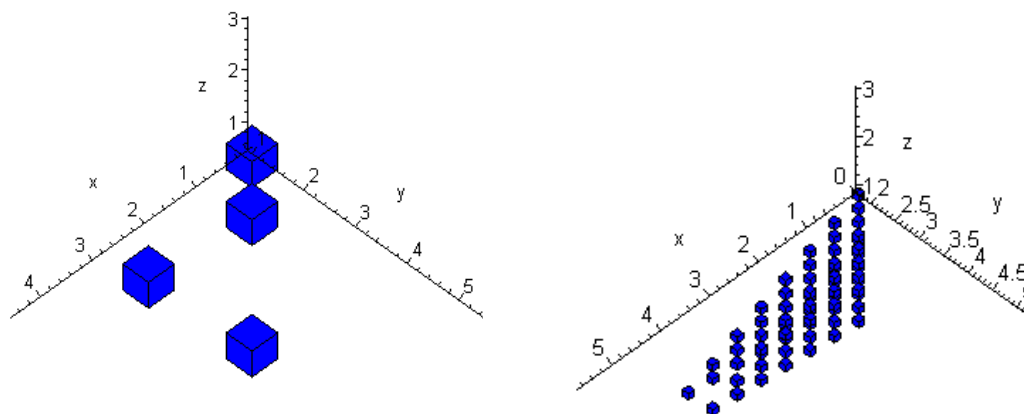
Taking into account [13] we get the form of IFS={F1,F2,F3,F4}, where  $F_i = P^{-1} \cdot B_i \cdot P$ ,  $i=1, \dots, 4$ . As the particular case of the latest IFS we can obtain another one, more convenient, form of IFS. Namely, IFS={F1,F2,F3,F4}  $F_i = P^{-1} \cdot B_i \cdot P, i = 1, \dots, 4$ . As the particular case of the latest IFS we can obtain another one, more convenient, form of IFS. Namely, IFS={F1,F2,F3,F4}, where

$$F_1 = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ \frac{x_1}{2} & \frac{y_1}{2} & \frac{z_1}{2} & 1 \end{bmatrix}, \quad F_2 = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ \frac{x_2}{2} & \frac{y_2}{2} & \frac{z_2}{2} & 1 \end{bmatrix}$$

$$F_3 = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & 0 \\ \frac{x_3}{2} & \frac{y_3}{2} & \frac{z_3}{2} & 1 \end{bmatrix}$$

$$F_4 = \begin{bmatrix} -\frac{1}{2} & 0 & 0 & 0 \\ 0 & -\frac{1}{2} & 0 & 0 \\ \frac{x_1 + x_2 + x_3}{2} & \frac{y_1 + y_2 + y_3}{2} & -\frac{1}{2} & 0 \\ \frac{z_1 + z_2 + z_3}{2} & & & 1 \end{bmatrix}$$

The above given form of IFS can be obtained immediately using subdivision matrices, if the fourth rows of zeros in  $B_i, i=1, \dots, 4$ , will be replaced by  $[1/2, 0, 0, 1/2]$ ,  $[0, 0, 1/2, 1/2]$ ,  $[0, 1/2, 0, 1/2]$ ,  $[1/2, 1/2, 1/2, 1/2]$ , respectively. It means that IFSs not determined uniquely. For different IFSs it is easily to observe different convergence to the attractor. IFSs given above generate attractor in  $R^4$ . After its projection on  $R^3$  we obtain the fractal triangle. In Fig.2 the fractal rendering process of a triangle is presented. The triangle appears in the progressive way.



*Fig. 2. A triangle rendered fractally using deterministic method starting from a box (iterations: 1,3,6).*

### **5- An Minimization Problem Using Least Square Method**

The minimization is due to the application of the discrete least square method which is the difference between the calculated and exact set of points constituting the attractor of the IFS, this will be done on minimizing the Hausdorff distance between these two sets [3].

Considering that, the fractal shape is given in advance, and the problem is to find the affine maps that constructing the IFS, the procedure is as follows:

- i. Select n number of data points from the fractal rendering shape after scaling the figure using a prespecified coordinate axis and arrange the obtained data points into ascending order.
- ii. Introducing an initial guess to the coefficients corresponding to the fractal rendering patches.
- iii. Applying the direct problem. The results of this program will produce a different shape, but has the property that their results depend on the initial parameters obtained from the inverse program.
- iv. Select, also n-number of data points resulted by applying the direct problem on the produced results of the inverse problem and rearranging these points ascending order.
- v. Evaluating the objective function using the discrete least square method. This function is given by:

$$d(A_e, A_a) = \sum_{i=1}^n (x_{e_i} - x_{a_i})^2$$

Where  $x_{e_i}$  is the x-coordinate of the original fractal shape, and  $x_{a_i}$  is the x-coordinate corresponding to the approximate fractal shape.

- vi. Minimize the objective function proposed in step (v) using Hooke and Jeeves [5] or any optimization method.

## **6- Conclusion Remark**

Due to the Simplest of direct optimization problem especial Hook & Jeeves method, the approximate method of “Hook & Jeeves” [5] will be used here in this work to find the numerical solution of the inverse problem..

## **8- Future Work**

- 1) Fractal dimension has been used to characterize data texture in a large number of physical and biological sciences. In medical imaging, Alan and Murray are presented in [2] a new method of estimating fractal dimension which is applicable to data-limited medical images and which shows promise of improving the reparability classes of images. One of the open future work to find relation ship between dimension of image in this method and the dimension of fractal shape in our work.
- 2) Using this subject in coding theory, so as to use a fractal shape as a password to the systems.

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