

The curvature and dimension of a closed surface

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Abstract

In this short memorandum, the curvature and dimension properties of the 2-sphere surface of a 3-dimensional ball and the $2.x$ -dimensional surface of a 3-dimensional fractal set are considered. Tessellation is used to approximate each surface, primarily because the $2.x$ -dimensional surface of a 3-dimensional fractal set is otherwise non-differentiable (having no well-defined surface normals). It is found that the curvature of a closed surface *must* lead to fractional dimension.

1 Overview

Unlike in traditional geometry where dimension is an integer (e.g. $(3+1)$ -dimensional space-time), fractional (non-integer) dimension occurs in *fractal* geometry. In fractal geometry, there are currently many ways to calculate the dimension of a surface [1, 2]. This memo uses a new method of calculating the fractional dimension of a surface – it is *curvature* that leads to this fractional dimension.

Our main focus will be on the curvature and dimension of tessellated closed surfaces. For example, Marching Cubes [3] can be used to generate triangular tessellations (meshes), where dimension $D \in (2.0, 3.0)$.

Our attention will be drawn to the difference in curvature and dimension between a 2-sphere and the $2.x$ -dimensional surface of a 3-dimensional fractal set. We will generate both a 2-sphere and the $2.x$ -dimensional surface of a 3-dimensional fractal set by using iterative quaternion equations.

Some notes are given at the end of this memo.

2 The tessellation of a closed surface

Approximating the surface of a 3-dimensional shape as a mesh allows us to calculate the surface's dimension $D \in (2.0, 3.0)$. This includes approximation of both a 2-sphere and the $2.x$ -dimensional surface of a 3-dimensional fractal set.

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First we calculate, for each triangle, the average dot product of the triangle's face normal \hat{n}_i and its 3 neighbouring triangles' face normals $\hat{o}_1, \hat{o}_2, \hat{o}_3$:

$$d_i = \frac{\hat{n}_i \cdot \hat{o}_1 + \hat{n}_i \cdot \hat{o}_2 + \hat{n}_i \cdot \hat{o}_3}{3} \in (-1.0, 1.0]. \quad (1)$$

Because we assume that there are 3 neighbours per triangle, the mesh must be *closed* (no cracks or holes, precisely 2 triangles per edge). The reason why the value -1.0 is not achievable is because that would lead to intersecting triangles.

Then we calculate the normalized measure of curvature, where A_i is the triangle area, and A_{largest} is the largest triangle area in the mesh:

$$k_i = \left(\frac{1 - d_i}{2} \right) \left(\frac{A_i}{A_{\text{largest}}} \right) \in [0.0, 1.0). \quad (2)$$

The triangle area is used like this because there are sliver triangles produced by Marching Cubes.

Once k_i has been calculated for all triangles, we can then calculate the average normalized measure of curvature K , where t is the number of triangles in the mesh:

$$K = \frac{1}{t} \sum_{i=1}^t k_i = \frac{k_1 + k_2 + \dots + k_t}{t} \in (0.0, 1.0). \quad (3)$$

The reason why the value 0.0 is not achievable is because we are dealing with a closed surface, and so there's bound to be *some* curvature.

The dimension of the closed surface is:

$$D = 2 + K \in (2.0, 3.0). \quad (4)$$

As far as we know, this method of calculating the dimension of a closed surface is new. The entire C++ code for generating a mesh can be found at [4]. The entire C++ code for calculating a mesh's curvature-based dimension can be found at [5].

3 Vanishing versus non-vanishing curvature

Where $r \in [2, \infty)$ is the *integer* sampling resolution, $g_{\text{max}} \in (-\infty, \infty)$ is the sampling grid maximum extent, $g_{\text{min}} \in (-\infty, \infty)$ is the sampling grid minimum extent, and $g_{\text{max}} > g_{\text{min}}$, the Marching Cubes step size is:

$$\ell = \frac{g_{\text{max}} - g_{\text{min}}}{r - 1} \in (0.0, 1.0). \quad (5)$$

In this memo $g_{\text{max}} = 1.5$, $g_{\text{min}} = -1.5$, and r is variable. Although ℓ can go up to almost infinity, we shall enforce a virtual cap at almost unity, as is traditionally done in the research.

On one hand, a 2-sphere can be generated by the iterative quaternion Julia set equation

$$Z = Z^2 + C, \quad (6)$$

where the translation constant is $C = 0.0, 0.0, 0.0, 0.0$. For a 2-sphere, the *local* curvature all but vanishes as ℓ decreases (as r increases):

$$\lim_{\ell \rightarrow 0.0} K(\ell) = 0.0. \quad (7)$$

This results in a dimension of practically (but never quite) 2.0, which is to be expected from a non-fractal surface. See Figures 1 - 3.

On the other hand, the $2.x$ -dimensional surface of a 3-dimensional fractal set can be generated by the iterative quaternion equation

$$Z = Z \cos(Z). \quad (8)$$

For the $2.x$ -dimensional surface of a 3-dimensional fractal set, the local curvature does not necessarily vanish as ℓ decreases:

$$\lim_{\ell \rightarrow 0.0} K(\ell) \neq 0.0. \quad (9)$$

This results in a dimension considerably greater than 2.0 (but not equal to or greater than 3.0), which is to be expected from a fractal surface. See Figures 4 - 7.

For more information on iterative quaternion equations, and how to perform quaternion multiplication, addition, and cos, see [6].

4 Notes

4.1 Box-counting dimension for the $2.x$ -dimensional surface of a 3-dimensional fractal set

One may wish to compare the curvature-based dimension given in this memo against the traditional box-counting (Minkowski-Bouligand) dimension. Thanks to Marching Cubes, it is very simple to obtain the number of boxes that are required to cover the $2.x$ -dimensional surface of a 3-dimensional fractal set. One can simply ask: how many triangles are generated per this or that marched cube? If the answer is greater than zero, then that particular marched cube covers the surface – the *integer* box count N is increased by one. The box-counting dimension D_{box} is calculated as usual, and is implemented in the mesh generation source code [4]:

$$D_{\text{box}} = \lim_{\ell \rightarrow 0.0} \frac{\log(N)}{\log(1/\ell)} \in [0.0, \infty), \quad (10)$$

where, again, $0.0 < \ell < 1.0$.

4.2 Comparing Marching Squares to Marching Cubes

Marching Squares [7] can be used to generate closed line paths, where dimension $D \in (1.0, 2.0)$. For each line segment, the average dot product of the line segment's surface normal \hat{n}_i and its 2 neighbouring line segments' surface normals \hat{o}_1, \hat{o}_2 is:

$$d_i = \frac{\hat{n}_i \cdot \hat{o}_1 + \hat{n}_i \cdot \hat{o}_2}{2} \in (-1.0, 1.0]. \quad (11)$$

Because we assume that there are 2 neighbours per line segment, the line segment mesh must be *closed* (precisely 2 line segments per end point). The reason why the value -1.0 is not achievable is because that would lead to intersecting line segments.

See Figure 8 for an example input (a 2-dimensional greyscale image, consisting of 8×8 pixels) and output (a $1.x$ -dimensional closed set of line segments) of the Marching Squares algorithm, approximating a 1-sphere (a circle), where sampling resolution is $r = 8$. Note that for Marching Cubes, the input is a 3-dimensional ‘greyscale image’, consisting of voxels, and the output is a $2.x$ -dimensional closed set of triangles.

Illustrated in Figure 9 is a section of a closed line path, with surface normals. The average dot product of neighbouring line segments is $d_i = 0.0$. This leads to a normalized measure of curvature $k_i = 0.5$, which in turn leads to an average normalized measure of curvature $K = 0.5$. The dimension is $D = 1 + K = 1.5$. Note that for Marching Cubes, the dimension is $D = 2 + K$.

See Figure 10 for a section of a closed line path as it goes from dimension 1.0 (at top) to 1.9999 (at bottom). In the end, where the dimension is 1.9999, the result is practically a rectangle. The reason why the dimension cannot be 2.0 is because that would lead to intersecting line segments. Note that for Marching Cubes, the dimension cannot be 3.0 because that would lead to intersecting triangles.

4.3 Introduction to Marching Hypercubes

There is research on Marching Hypercubes at [8, 9, 10]: where dimension $D \in (3.0, 4.0)$, the output can be a closed set of tetrahedra. For example, where P_{tc} is the i th tetrahedron centre, P_{fc} is the first face’s centre, and P_{ntc} is the first neighbouring tetrahedron’s centre:

$$\hat{n}_1 = \text{normalize}(P_{tc} - P_{fc}), \quad (12)$$

$$\hat{o}_1 = \text{normalize}(P_{fc} - P_{ntc}), \quad (13)$$

and likewise for the other 3 neighbouring tetrahedra, the value of d_i is:

$$d_i = \frac{\hat{n}_1 \cdot \hat{o}_1 + \hat{n}_2 \cdot \hat{o}_2 + \hat{n}_3 \cdot \hat{o}_3 + \hat{n}_4 \cdot \hat{o}_4}{4} \in (-1.0, 1.0]. \quad (14)$$

Because we assume that there are 4 neighbours per tetrahedron, the tetrahedral mesh must be *closed* (precisely 2 tetrahedra per face). The reason why the value -1.0 is not achievable is because that would lead to intersecting tetrahedra.

Imagine wrinkly space, begetting curvature, and thus increased dimension. Space would technically be a ‘ $3.x$ -sphere’ – practically a 3-sphere, with the exception of pockets of local curvature that occur along said surface. This is not to say that in real life space is quantized into tetrahedra, like in [11], nor that the topology of the Universe is closed in real life. That said, if space *on the computer* is quantized as such, then there is a minimum non-zero curvature because there is no such thing as a tessellation via regular tetrahedra alone – irregularity, and thus curvature, are required. This curvature is above and beyond the minimum non-zero curvature that is due to the fact that a $3.x$ -sphere is closed, and thus inherently curved.

4.4 Acknowledgements

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Figure 1: Low resolution ($r = 10$) surface for the iterative quaternion equation $Z = Z^2 + C$, where $C = 0.0, 0.0, 0.0, 0.0$. The surface's dimension is 2.02.



Figure 2: Medium resolution ($r = 100$) surface for the iterative quaternion equation $Z = Z^2 + C$, where $C = 0.0, 0.0, 0.0, 0.0$. The surface's dimension is 2.06.



Figure 3: High resolution ($r = 1000$) surface for the iterative quaternion equation $Z = Z^2 + C$, where $C = 0.0, 0.0, 0.0, 0.0$. The surface's dimension is practically 2.0.



Figure 4: Low resolution ($r = 10$) surface for the iterative quaternion equation $Z = Z \cos(Z)$. The surface's dimension is 2.05.



Figure 5: Medium resolution ($r = 100$) surface for the iterative quaternion equation $Z = Z \cos(Z)$. The surface's dimension is 2.11.



Figure 6: High resolution ($r = 1000$) surface for the iterative quaternion equation $Z = Z \cos(Z)$. The surface's dimension is 2.08.



Figure 7: A 2-dimensional slice of the iterative quaternion equation $Z = Z \cos(Z)$, showing the self-similar nature of the set at all scales.



Figure 8: Marching Squares input and output.



Figure 9: $D = 1.5$.



Figure 10: $D = 1$ to $D = 1.9999$.