Drawing 3D Fractals in Real Time

An Introduction to Distance Estimators

Giovanni Viglietta JAIST – March 19, 2019



Drawing 3D Fractals in Real Time

Outline

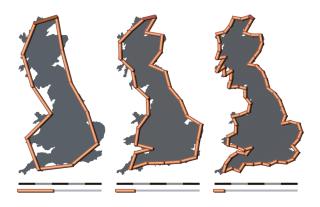
- Fractals in nature
- Fractal dimension
- The Mandelbrot set
- Rendering 3D objects
- Distance estimators
- Generalizing the complex numbers
- Rendering 3D Mandelbrot-like sets
- Open problems

The coastline paradox



The English mathematician Lewis F. Richardson (1881–1953), in his analysis of war, searched for a relation between the probability of two countries going to war and the length of their common border. However, in collecting data, he found considerable disagreements in the various published border lengths.

The coastline paradox



He then measured the length of the coast of Britain, and noticed that the length increases as the "ruler" gets shorter (e.g., 2300 km, 2800 km, 3500 km).

Can we take the <u>limit</u> as the real length? No: Richardson showed that the coast length diverges as the ruler gets shorter!

Similar paradoxes in nature



Many other objects in nature exhibit the same paradox: frost crystals, broccoli, the human brain...

Similar paradoxes in nature



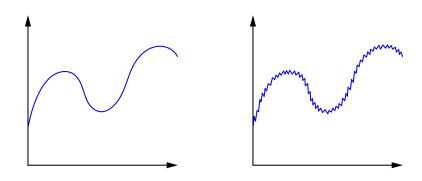
Many other objects in nature exhibit the same paradox: frost crystals, broccoli, the human brain...

Similar paradoxes in nature



Many other objects in nature exhibit the same paradox: frost crystals, broccoli, the human brain...

Traditional calculus



The central assumption of <u>traditional calculus</u> is that objects tend to look <u>smooth</u> if you zoom in far enough. This idealized notion does not capture the <u>roughness</u> of most real objects.



Between 1967 and 1975, Benoit Mandelbrot synthesized centuries of observations on undifferentiable, infinitely self-similar functions, formulating the notion of <u>fractal dimension</u>. It describes how a shape scales differently from the space in which it is embedded.

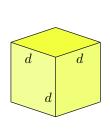


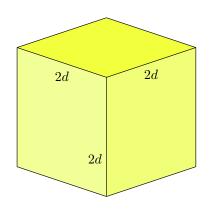
If a <u>disk</u> is scaled by a factor of 2, its area goes from πr^2 to $4\pi r^2$. Its fractal dimension is $\log_2 4=2$.



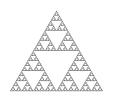
If a <u>circle</u> is scaled by 2, its length goes from $2\pi r$ to $2\cdot 2\pi r$.

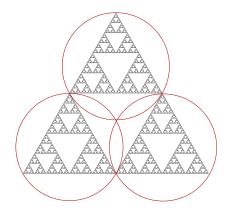
Its fractal dimension is $\log_2 2 = 1$.





If a <u>cube</u> is scaled by a factor of 2, its volume goes from d^3 to $8d^3$. Its fractal dimension is $\log_2 8 = 3$.

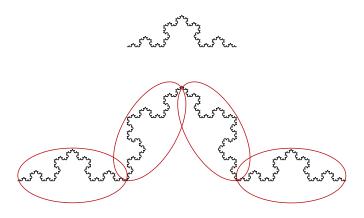




If a Sierpinski triangle is scaled by 2, its area is scaled by 3.

Its fractal dimension is $\log_2 3 \approx 1.585$.

Its dimension is not an integer, hence it is a fractal!



If a Koch snowflake is scaled by 3, its length is scaled by 4.

Its fractal dimension is $\log_3 4 \approx 1.262$.

Its dimension is not an integer, hence it is a fractal!

Approximate fractal dimensions of real objects:

• Coast of Britain: ~ 1.25

• Coast of Norway: ~ 1.52

• Outline of clouds: ~ 1.6

• Surface of oceans: ~ 2.3

• Surface of broccoli: ~ 2.7

• Surface of human brain: ~ 2.79

ullet Surface of human lungs: ~ 2.97

For each complex number c, define the function $f_c(z) = z^2 + c$.

Consider the iterations

```
f_c(0),

f_c(f_c(0)),

f_c(f_c(f_c(0))),

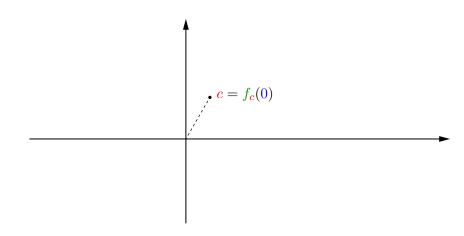
f_c(f_c(f_c(f_c(0)))),

\vdots

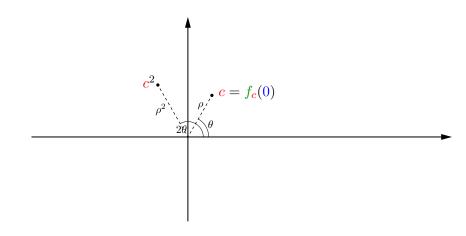
f_c^n(0).
```

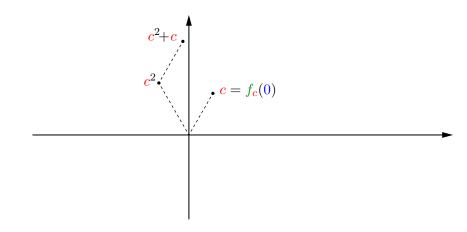
Definition

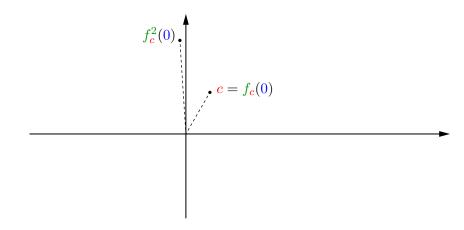
The complex number c is in the Mandelbrot set if and only if the sequence of moduli $|f_c^n(0)|$ is bounded.

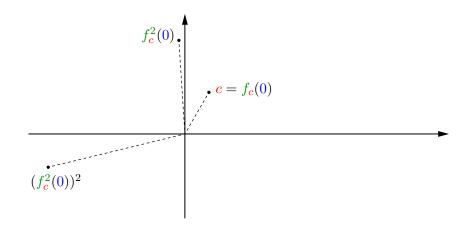


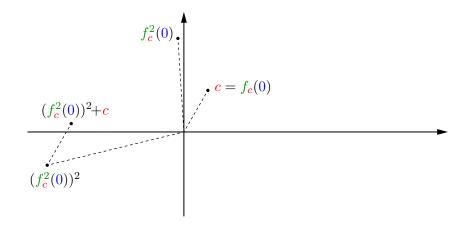
Is c in the Mandelbrot set?

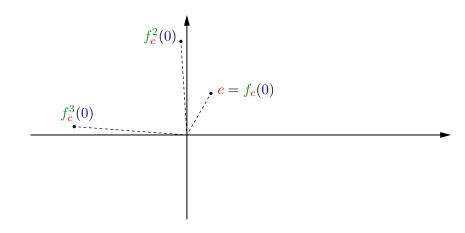


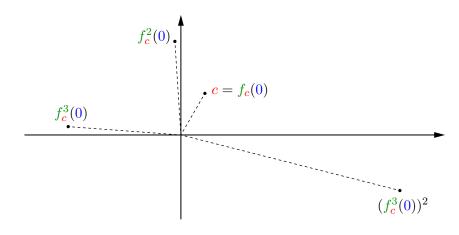


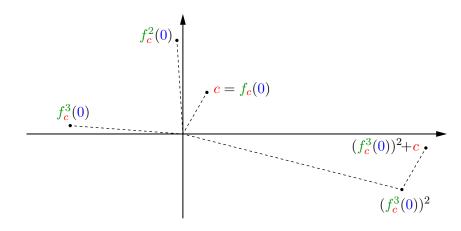


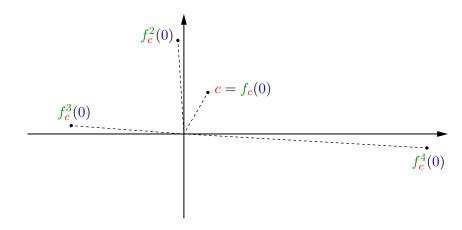










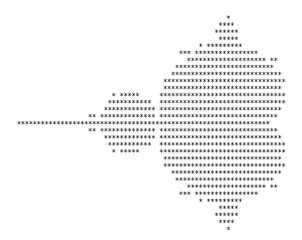


The moduli $|f_c^n(0)|$ are going to diverge, hence c is not in the Mandelbrot set.

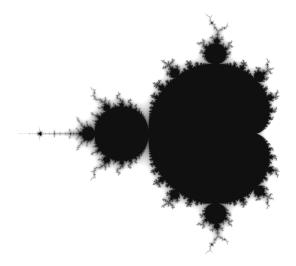
Is c = -1 in the Mandelbrot set?

Iterate
$$f_{-1}(z) = z^2 - 1$$
:
 $f_{-1}(0) = -1$,
 $f_{-1}(f_{-1}(0)) = 0$,
 $f_{-1}(f_{-1}(f_{-1}(0))) = -1$,
 $f_{-1}(f_{-1}(f_{-1}(f_{-1}(0)))) = 0$,
 \vdots

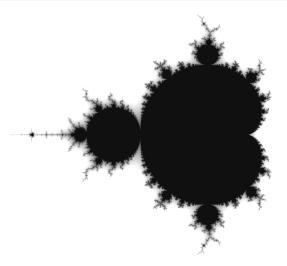
The moduli $|f_{-1}^n(0)|$ are <u>bounded</u>, hence -1 is in the Mandelbrot set.



The first published picture of the Mandelbrot set, by Robert W. Brooks and Peter Matelski in 1978.



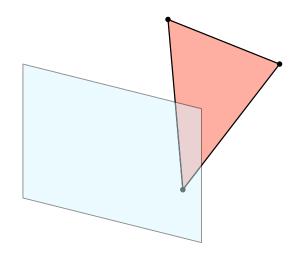
Increasing the resolution, a complex shape with infinitely self-similar features appears...



Fun fact: in 1998, Mitsuhiro Shishikura proved that the fractal dimension of the Mandelbrot set is 2, hence it is not a fractal!

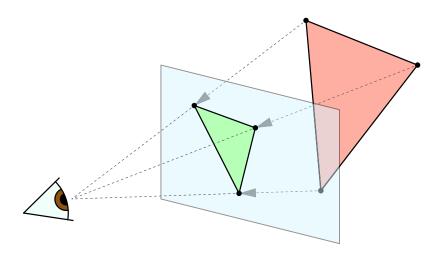
Main rendering techniques:

- Rasterization
- 2
- 3

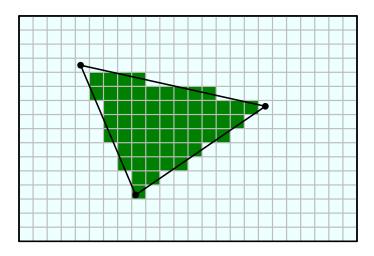




A <u>3D scene</u> has to be represented on a <u>2D screen</u> as seen from an observer positioned behind the screen.



In the <u>rasterization</u> technique, the vertices of a 3D shape are first projected onto the screen...



...then, the rasterizer identifies the <u>pixels</u> that fall into the shape, and colors them appropriately.



Rasterization has always been the most popular technique. Video games and <u>real-time</u> applications heavily rely on it.

Main rendering techniques:

Rasterization

Not suitable for drawing fractals (infinitely many vertices!)

2

(3

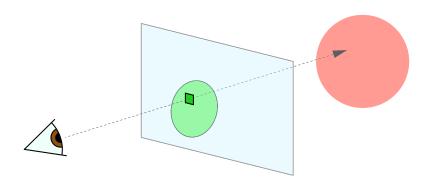
Main rendering techniques:

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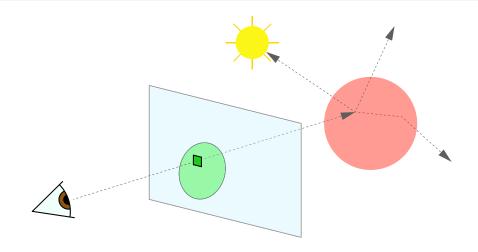
Not suitable for drawing fractals (infinitely many vertices!)

Ray tracing

3



In the <u>ray tracing</u> technique, a ray is cast for each pixel on screen, and the intersection with the closest object is computed.



Additional rays may be cast to compute shadows, reflections, refractions, caustics, etc.



Ray tracing gives the best results, but is very <u>slow</u>. This is the preferred technique for <u>animated movies</u>.

Main rendering techniques:

Rasterization

Not suitable for drawing fractals (infinitely many vertices!)

Ray tracing

Not suitable for real-time rendering (heavy computations!)

3

Main rendering techniques:

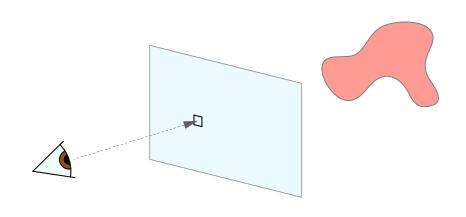
Rasterization

Not suitable for drawing fractals (infinitely many vertices!)

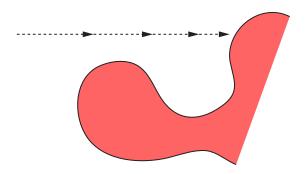
Ray tracing

Not suitable for real-time rendering (heavy computations!)

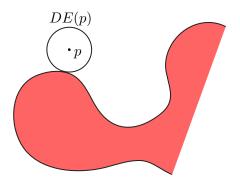
Ray marching



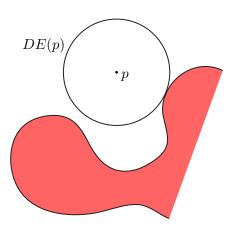
Ray marching starts like ray tracing: a ray is cast for each pixel on screen.



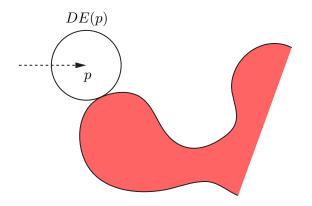
But instead of computing the exact intersection with a shape, it finds better and better approximations of it.

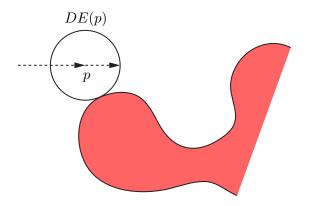


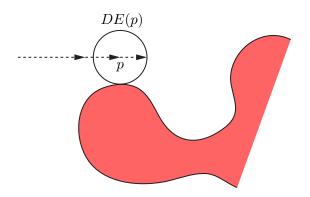
Suppose we are given a <u>distance estimator</u> for the shape: a function DE that tells the distance of each point from the shape.

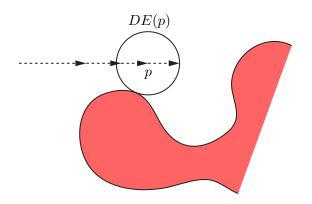


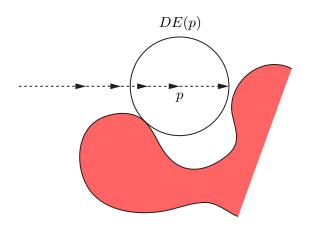
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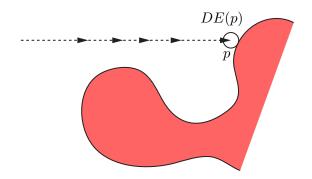




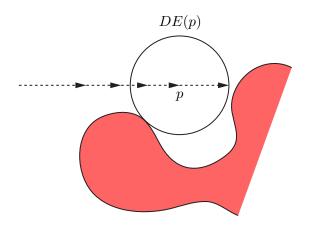




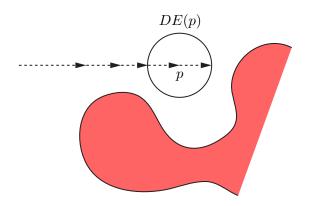




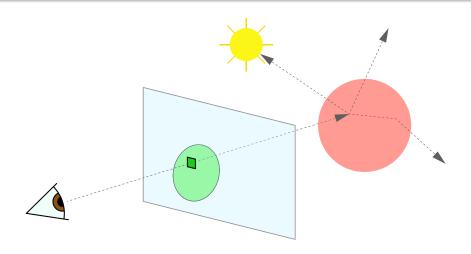
As soon as DE(p) is smaller than some <u>threshold</u>, we claim that the ray has hit the shape, and we stop.



Note that DE(p) does not have to be the exact distance, but it could be a <u>lower bound</u> (hence the name "estimator").

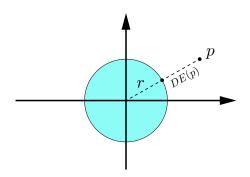


Note that DE(p) does not have to be the exact distance, but it could be a <u>lower bound</u> (hence the name "estimator").



As with ray tracing, <u>additional rays</u> may be cast to compute shadows, reflections, refractions, caustics, etc.

Distance estimator of a sphere

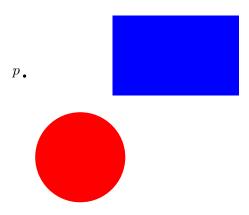


An exact distance estimator for a sphere of radius r is easy to compute: DE(p) = ||p|| - r.

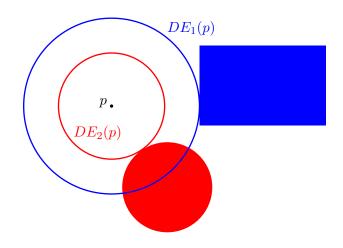
More distance estimators

Exact distance estimators for some simple shapes:

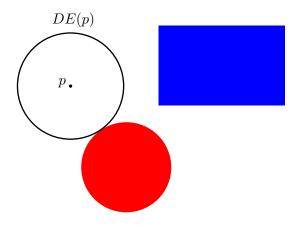
- Sphere: DE(p) = ||p|| r
- Cylinder: $DE(p) = ||p_{xy}|| r$
- Cone: $DE(p) = (\|p_{xy}\|, p_z) \cdot c$
- Torus: $DE(p) = \|(\|p_{xy}\| r_1, p_z)\| r_2$
- Plane: $DE(p) = p \cdot n + d$
- Box: $DE(p) = \|\max\{|p| b, 0\}\|$



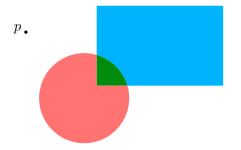
What if there is more than one shape in the scene?



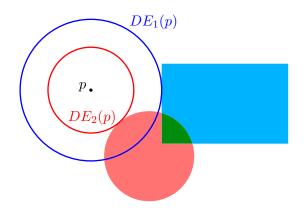
Suppose we have a distance estimator for each shape.



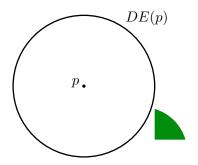
The combined distance estimator is the minimum.



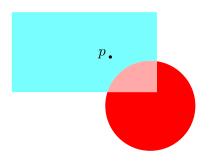
Similarly, the distance estimator of the <u>intersection</u> is the maximum.



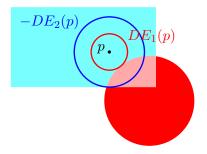
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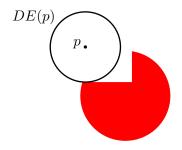
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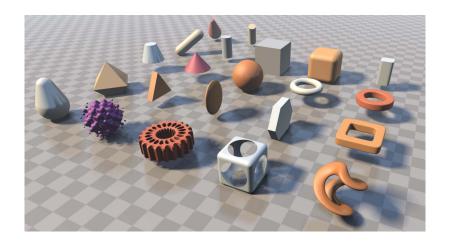
What if we want to <u>subtract</u> one shape from another?



Negate the distance estimator of the shape to be subtracted, which gives a distance estimator for its complement shape.



Then compute the <u>maximum</u>, as we did for the <u>intersection</u> of shapes.





Infinite repetition



We already know the distance estimator for a sphere:

$$DE(p) = ||p|| - r.$$

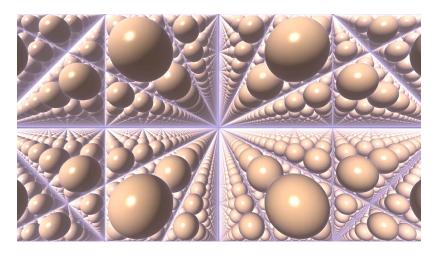
Infinite repetition



Let us make a small modification, introducing a mod operator:

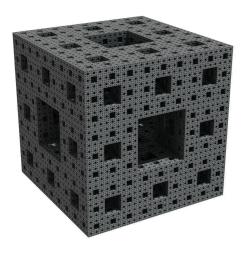
$$DE(p) = ||(p \mod m) - m/2|| - r.$$

Infinite repetition



Suddenly we get <u>infinitely many</u> spheres, at roughly the same cost! This would be <u>unfeasible</u> with rasterization or ray tracing.

Scale-invariant fractals



With similar methods, it is very easy to efficiently draw scale-invariant fractals such as the Menger sponge.

Generalizing the Mandelbrot set

The Mandelbrot set is defined from the iterations of the function

$$f_{\mathbf{c}}(z) = z^2 + \mathbf{c},$$

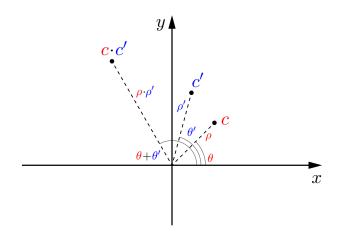
where z and c are complex numbers.

Is there a way to extend this notion to 3D?

- The term +c poses no problem (\mathbb{R}^3 is a vector space)
- But how should we interpret the term z^2 ?

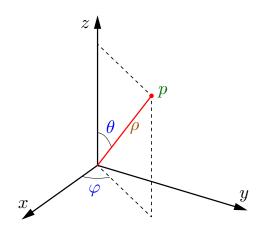
We need to generalize complex multiplication to some 3D space.

Generalizing complex multiplication



Recall the geometric interpretation of complex multiplication: in polar coordinates, multiply the norms and add the angles.

Generalizing complex multiplication



We can do the same thing in <u>spherical coordinates</u>, except that now we have two pairs of angles to add.

Distance estimator for the Mandelbrot set

To apply the <u>ray marching</u> technique to this 3D Mandelbrot set, we still need to define a <u>distance estimator</u> for it.

For the <u>2D Mandelbrot set</u>, an approximated distance estimator can be defined based on the theory of electrostatic potential.

At the <u>nth iteration</u> of $f_c(0)$, we get the complex number $g_n(c) = f_c^n(0)$. The distance estimator is:

$$DE(\mathbf{c}) = \lim_{n \to \infty} \frac{|g_n(\mathbf{c})| \cdot \ln |g_n(\mathbf{c})|}{2 \cdot |g'_n(\mathbf{c})|},$$

where $g'_n(c)$ is the complex derivative of $g_n(c)$ with respect to c.



Distance estimator for the Mandelbrot set

How do we compute the complex derivative of $g_n(c) = f_c^n(0)$?

From $f_c(z) = z^2 + c$, we have the recursion:

$$g_n(\mathbf{c}) = f_{\mathbf{c}}(g_{n-1}(\mathbf{c})) = (g_{n-1}(\mathbf{c}))^2 + \mathbf{c}.$$

Deriving with respect to c, we get:

$$g'_n(\mathbf{c}) = 2 \cdot g_{n-1}(\mathbf{c}) \cdot g'_{n-1}(\mathbf{c}) + 1 = 2 \cdot f_{\mathbf{c}}^{n-1}(\mathbf{0}) \cdot g'_{n-1}(\mathbf{c}) + 1.$$

Distance estimator for the Mandelbrot set

How do we compute the complex derivative of $g_n(c) = f_c^n(0)$?

From $f_c(z) = z^2 + c$, we have the recursion:

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Deriving with respect to c, we get:

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Can we extend the same idea to the 3D Mandelbrot set?

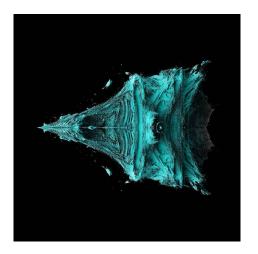
Problem 1: what is the potential of the 3D Mandelbrot set?

Tentative solution: use the <u>same</u> potential function as in 2D!

Problem 2: the complex derivative is a complex number, but for points with three components we would get a 3x3 <u>Jacobian matrix</u>.

Tentative solution: use the <u>same</u> recursive formula as in 2D, and take <u>norms</u> $||g'_n(c)||$ instead of moduli $|g'_n(c)|$

The Mandelbulb



The result is not so aesthetically pleasing and does not exhibit a lot of self-similarity.

The Mandelbulb



But if we replace $f_{\it c}(z)=z^2+{\it c}$ with $f_{\it c}(z)=z^8+{\it c}$, we obtain the Mandelbulb.

Another approach

Is there another way to extend complex multiplication to 3D?

Let us add to 1 and i a third component j. If we assume the distributive law, we only have to fill out this table:

×	1	i	j
1	1	i	j
i	i	-1	?
j	j	?	?

Another approach

Is there another way to extend complex multiplication to 3D?

Let us add to 1 and i a third component j. If we assume the distributive law, we only have to fill out this table:

×	1	i	j
1	1	i	\overline{j}
i	i	-1	?
j	j	?	?

Unfortunately, if we also want associativity, we get

$$i \cdot (\mathbf{i} \cdot \mathbf{j}) = (i \cdot i) \cdot j = -j,$$

but trying to solve the equation $i \cdot ? = -j$ always causes j to be a linear combination of 1 and i, making the space 2-dimensional.

There is no solution in 3D.

Quaternions

A possible workaround is using 4 components instead of 3, which yields the quaternions \mathbb{H} :

×	1	i	j	\mathbf{k}
1	1	i	j	\overline{k}
i	i	-1	k	-j
j k	j	-k	-1	i
\mathbf{k}	k	j	-i	-1

Quaternions

A possible workaround is using 4 components instead of 3, which yields the quaternions \mathbb{H} :

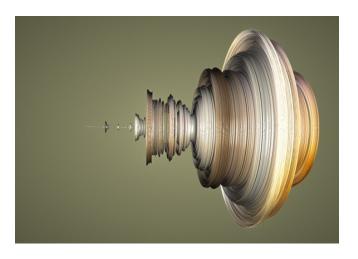
X	1	i	\mathbf{j}	\mathbf{k}
1	1	i	j	k
i	i	-1	k	-j
j k	j	-k	-1	i
\mathbf{k}	k	j	-i	-1

Theorem (Frobenius, 1877)

The only finite-dimensional vector spaces on the real numbers with associative and distributive multiplication and division are \mathbb{R} , \mathbb{C} , \mathbb{H} .

Problem: how do we visualize a 4D object?

Quaternion Mandelbrot set



If we keep the k component at a fixed value, we just get a solid of revolution.

Quaternion Mandelbrot set



But if we apply the transformation $p_y := -p_y$ at each iteration, we obtain a more interesting shape.

Quaternion Mandelbrot-like sets



By adding more parameters to the iteration function, we get higher-dimensional shapes that can be "sliced" in various ways.



Alan Norton

Generation and display of geometric fractals in 3-D SIGGRAPH '82

We illustrated the <u>ray marching</u> technique with distance estimators for real-time rendering of <u>3D fractals</u>.

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- Design video games around these concepts: navigation and interaction with <u>fractal environments</u> at various scales, etc.

