

# Advanced Mathematical Perspectives 1

## Lecture 21: Fractals and Statistical Self-Similarity



Matthew Roughan

[<matthew.roughan@adelaide.edu.au>](mailto:matthew.roughan@adelaide.edu.au)

[www.maths.adelaide.edu.au/matthew.roughan/notes/AMP1/](http://www.maths.adelaide.edu.au/matthew.roughan/notes/AMP1/)

School of Mathematical Sciences,  
University of Adelaide



THE UNIVERSITY  
of ADELAIDE



**ACEMS**

AUSTRALIAN RESEARCH COUNCIL CENTRE OF EXCELLENCE FOR  
MATHEMATICAL AND STATISTICAL FRONTIERS

So, Nat'ralists observe, a flea  
Hath smaller fleas that on him prey;  
And these have smaller still to bite 'em  
And so proceed ad infinitum  
*Jonathon Swift, 1733*

Why is geometry often described as “cold” and “dry?” One reason lies in its inability to describe the shape of a cloud, a mountain, a coastline, or a tree. Clouds are not spheres, mountains are not cones, coastlines are not circles, and bark is not smooth, nor does lightning travel in a straight line. Nature exhibits not simply a higher degree but an altogether different level of complexity.

*Benoit Mandelbrot*

# Section 1

## Fractals

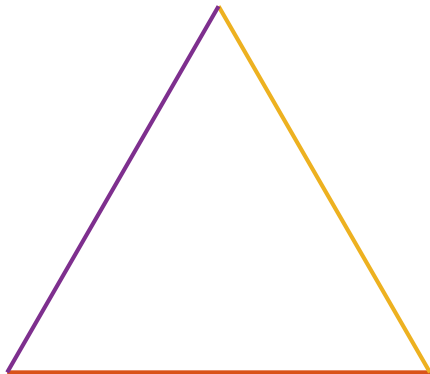
# The first (deterministic) fractal (1883)

## The Cantor set

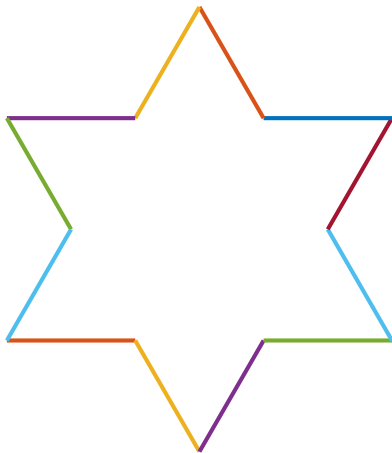


- Built by repeating a rule, so that the result looks the same under scaling transformations, *i.e.*, it has a symmetry called scale invariance or *self-similarity*
- But Cantor didn't call it a fractal

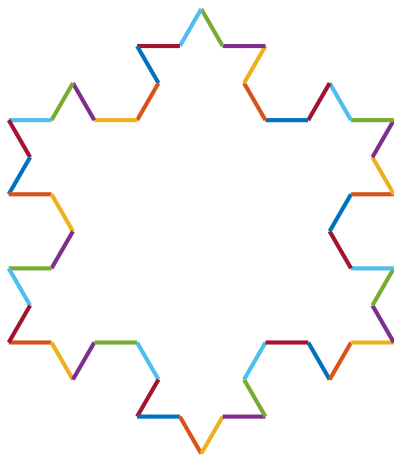
# Self-similarity: von Koch Snowflake (1904)



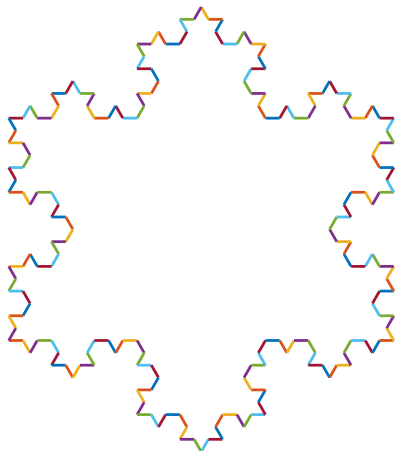
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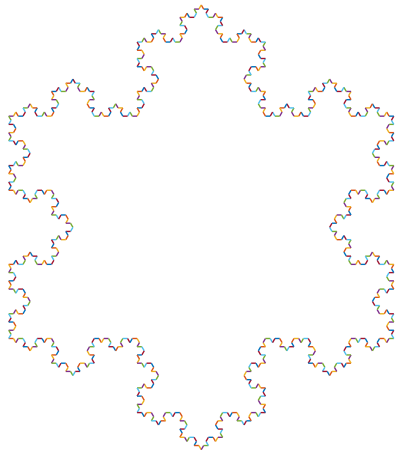


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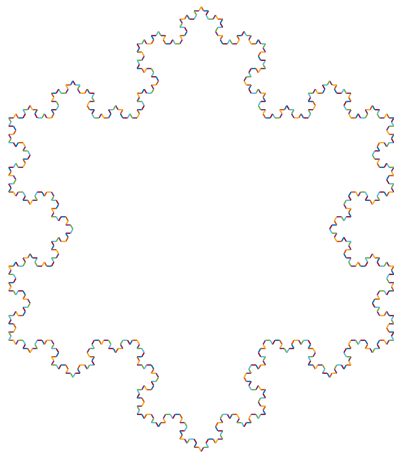




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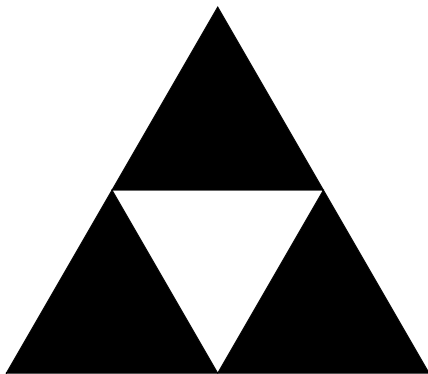
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And on *ad infinitum*

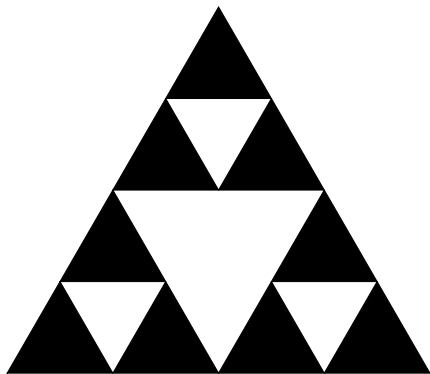
# Sierpinski Gasket or Triangle

The Sierpinski Gasket appeared c13th century, long before was “discovered” [CT11]



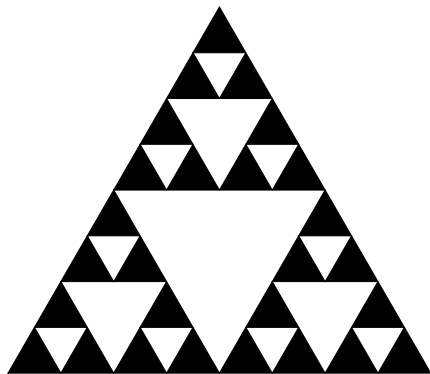
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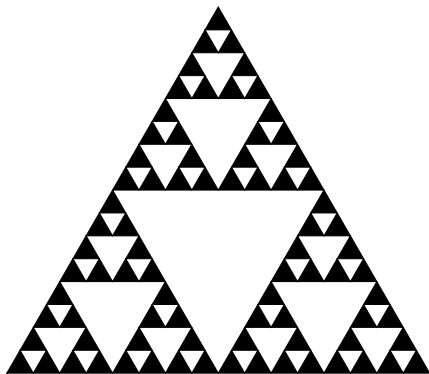
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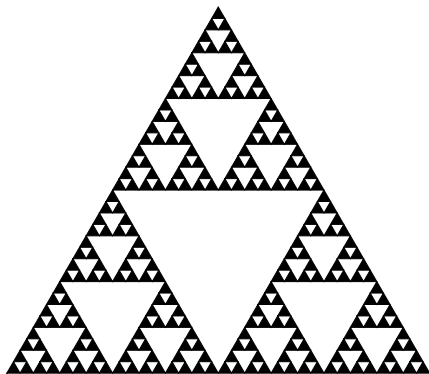
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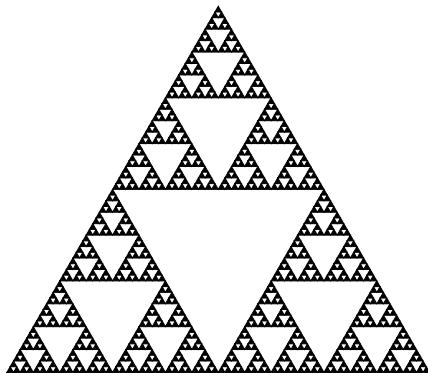
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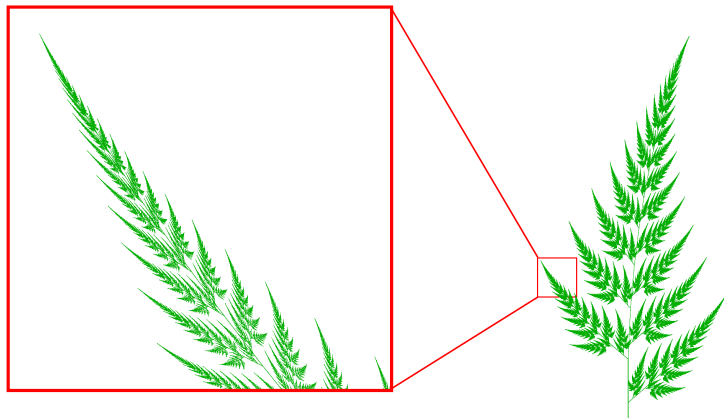
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And on *ad infinitum* ↻



# Self-similarity: Iterated Function System (IFS) Fern



C code from

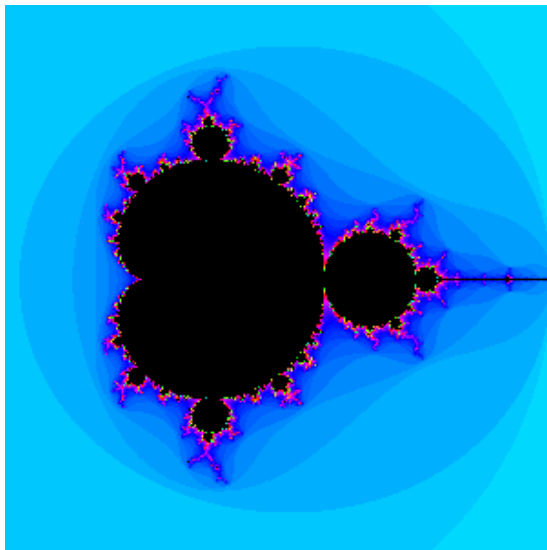
<http://astronomy.swin.edu.au/~pbourke/fractals/>

But the link is dead, so guess you should head here:

<http://paulbourke.net/fractals/>

# Mandelbrot Set

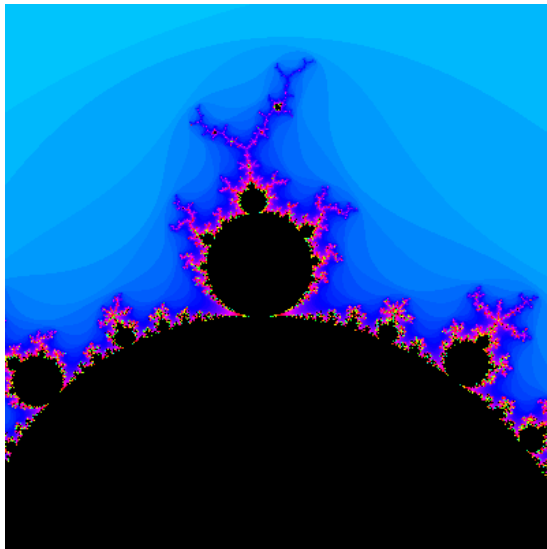
Fractals aren't defined in terms of strict self-similarity



<http://aleph0.clarku.edu/~djoyce/julia/julia.html>

# Mandelbrot Set

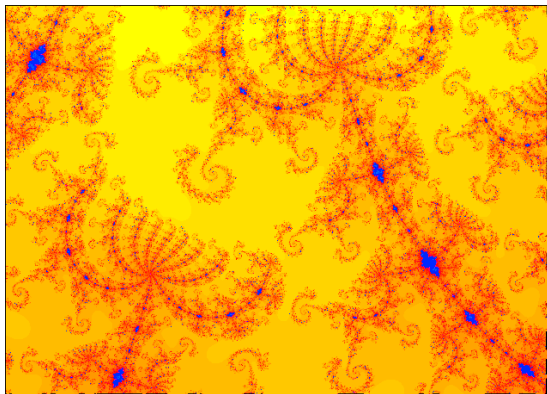
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# Mandelbrot Set

Fractals aren't defined in terms of strict self-similarity



<http://www.softsource.com/softsource/fractal.html>

# Mandelbrot Set DIY

Take the set of functions:  $f_c : \mathbb{C} \rightarrow \mathbb{C}$  for  $c \in \mathbb{C}$ , defined by

$$f_c(z) = z^2 + c$$

The **Mandelbrot set** is

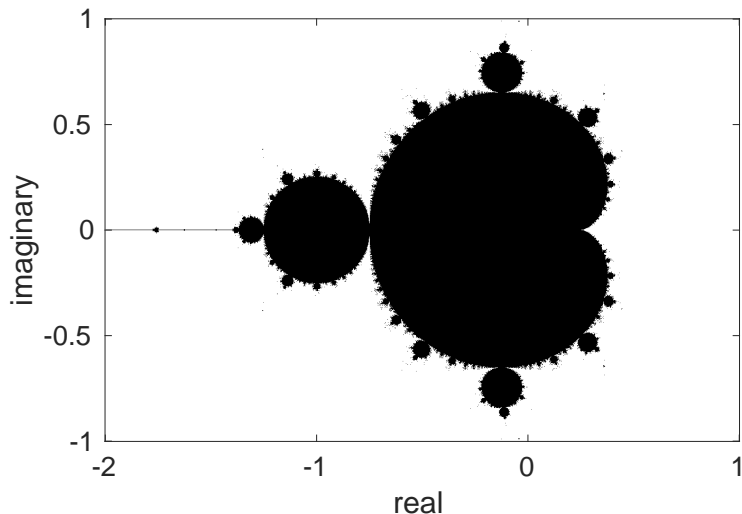
$$M = \left\{ c \in \mathbb{C} \mid \sup_{n \in \mathbb{N}} |f_c^n(0)| < \infty \right\}$$

- starting point  $z_0 = 0$
- for each point  $c$  in the complex plane, iterate  $f_c(\cdot)$
- if it diverges, it's not in the set

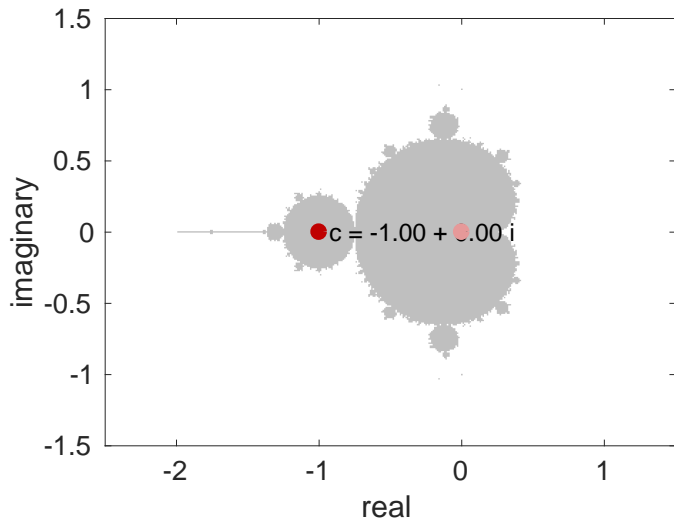
That defines the “set”, but usually people plot colours on the points depending on how quickly they diverge.

# Mandelbrot "Set"

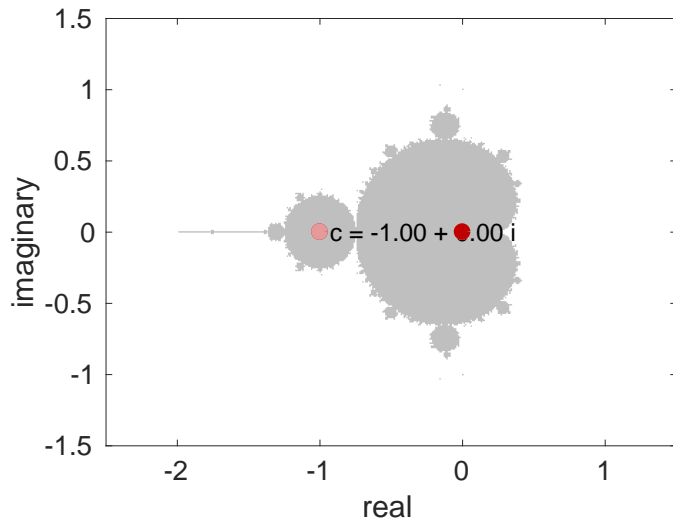
Axis refer to real and imaginary parts of  $c$ , black points are in the set.



# Mandelbrot Set Example Sequence I

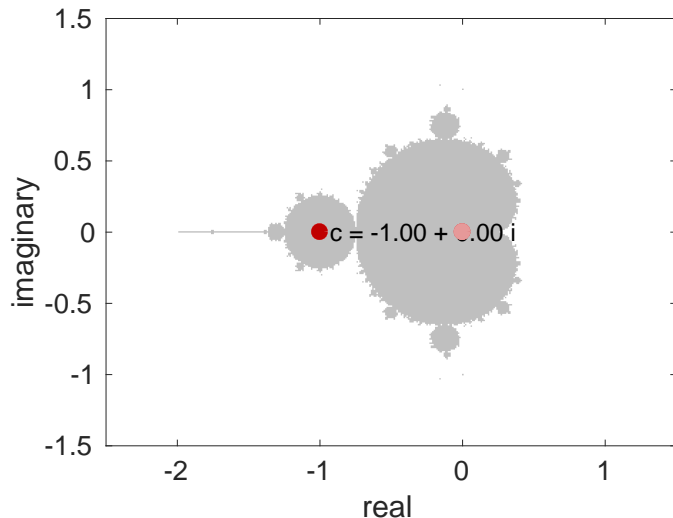


# Mandelbrot Set Example Sequence I

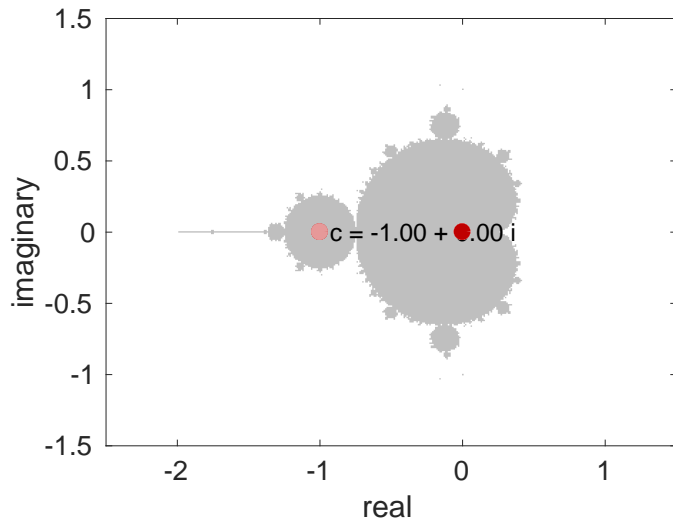




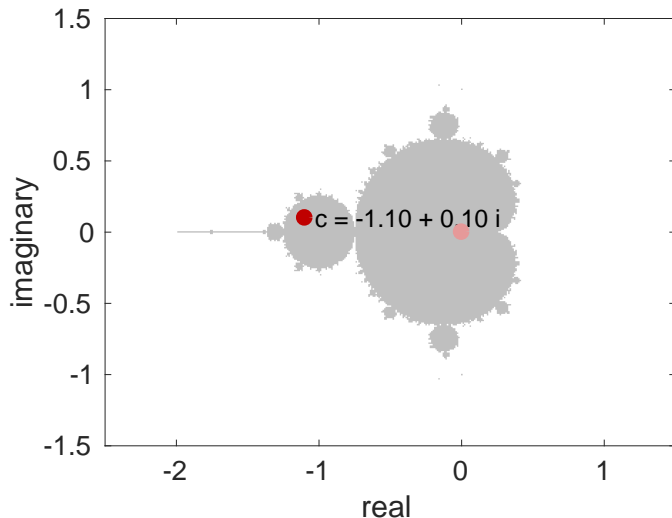
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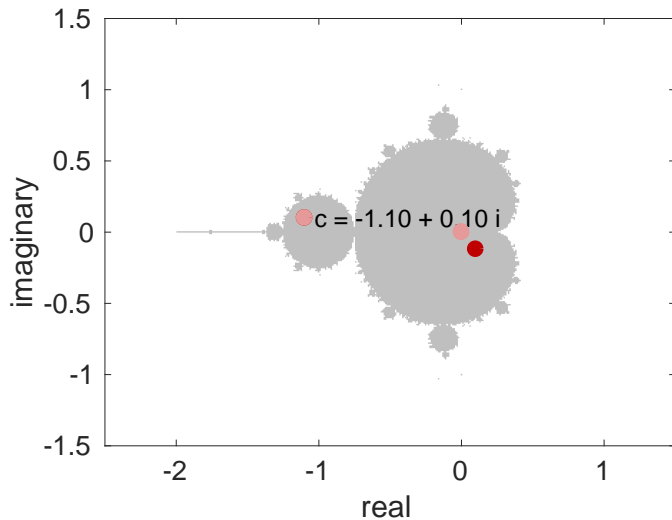
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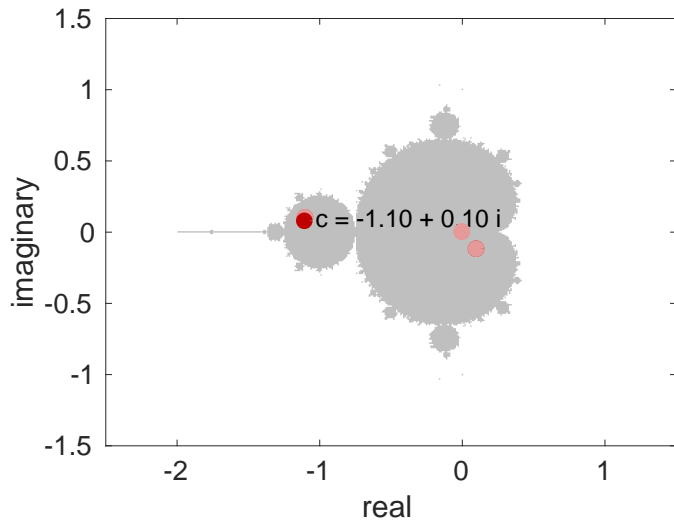
## Mandelbrot Set Example Sequence II



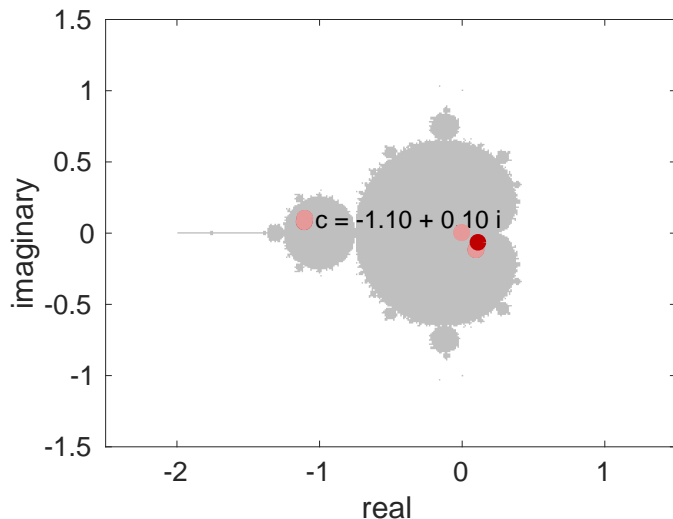
## Mandelbrot Set Example Sequence II



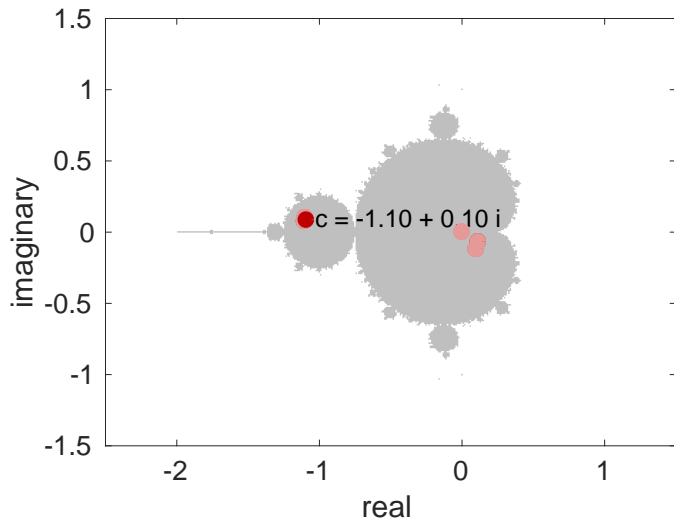
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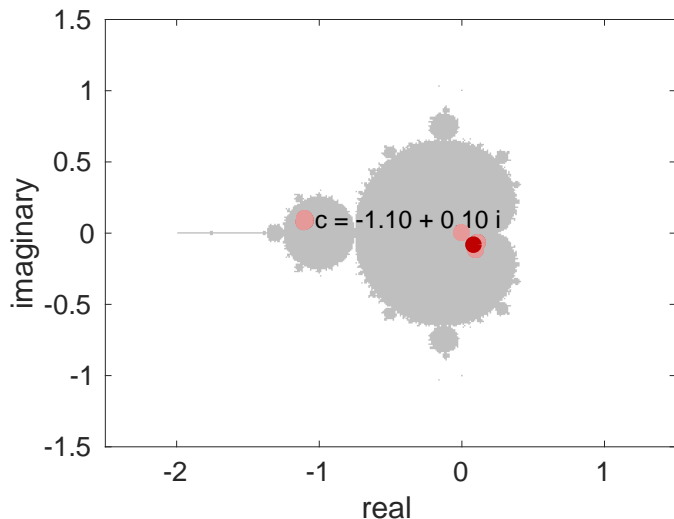
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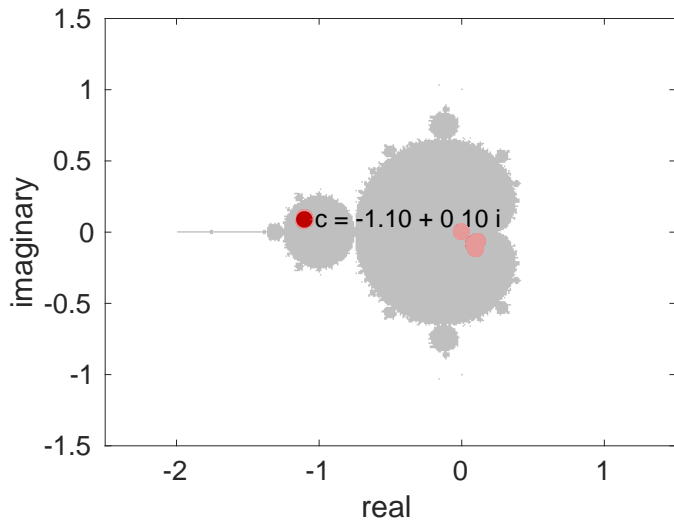


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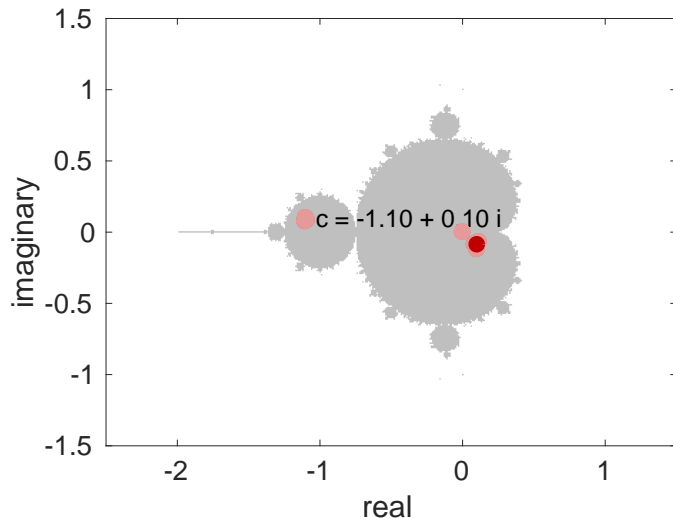




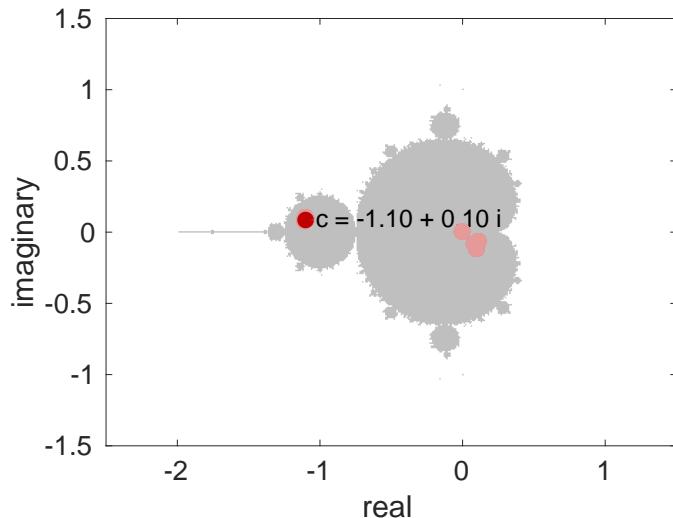
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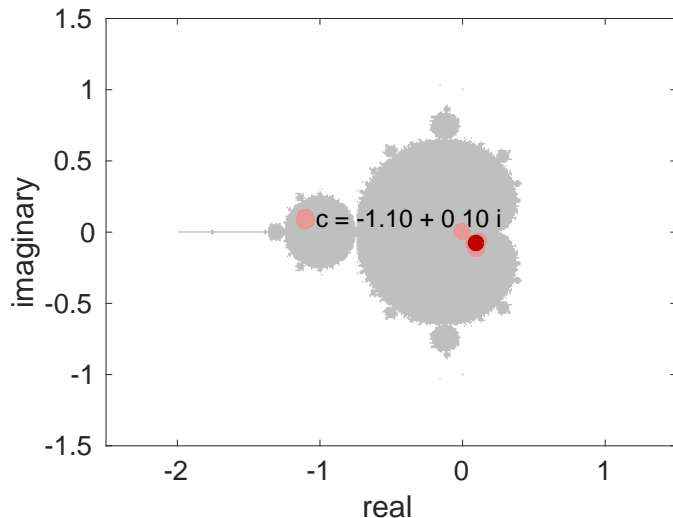
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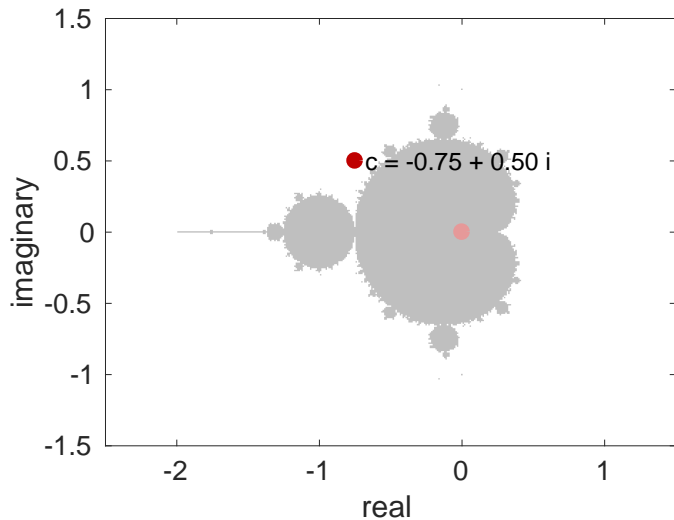
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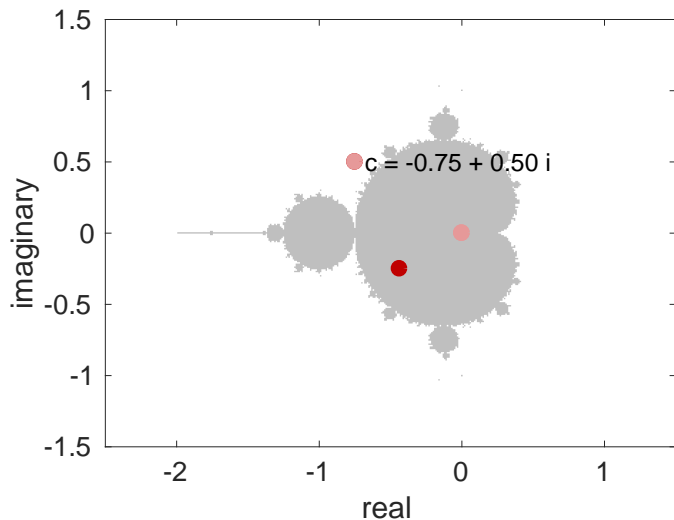
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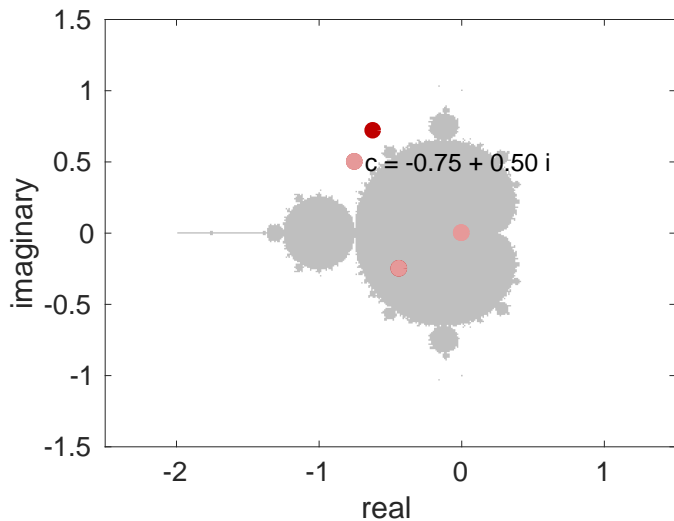
# Mandelbrot Set Example Sequence III



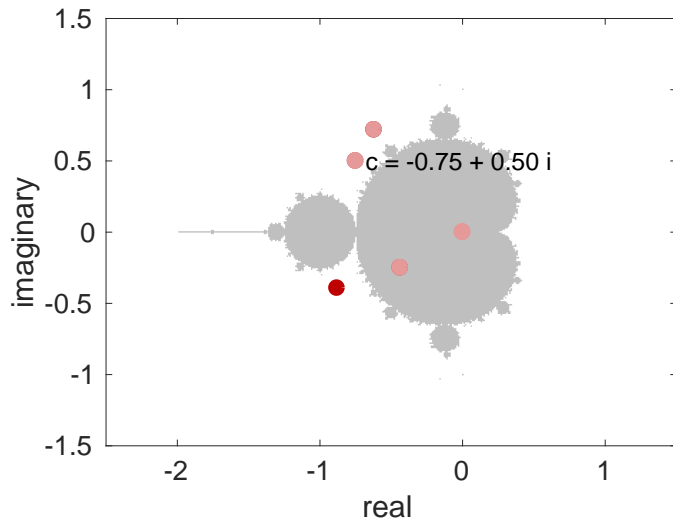
# Mandelbrot Set Example Sequence III



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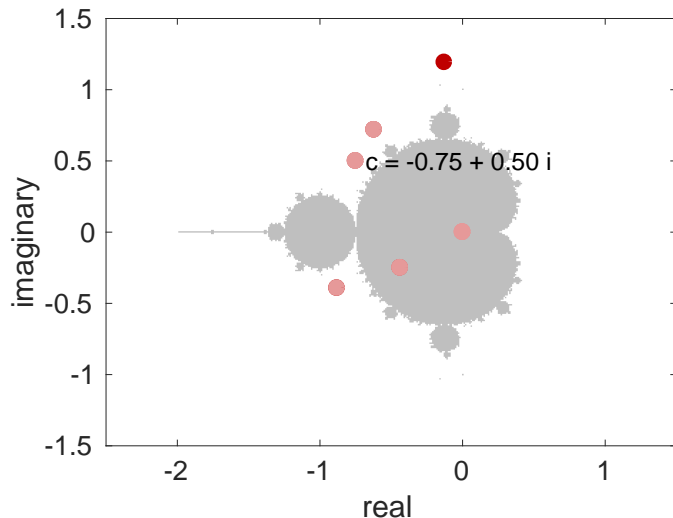


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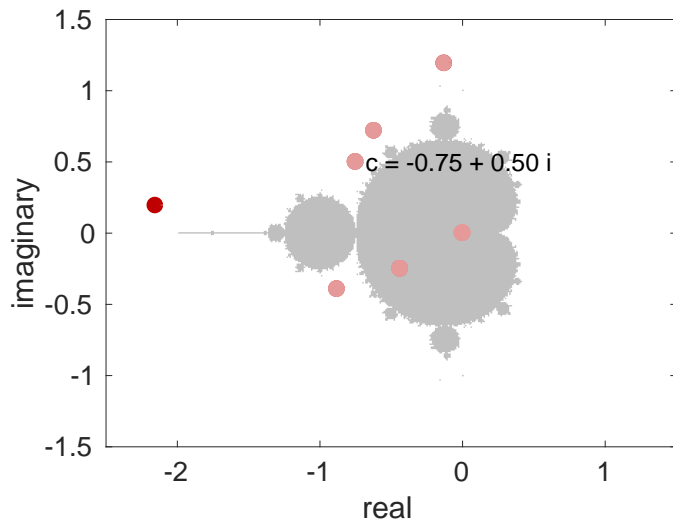




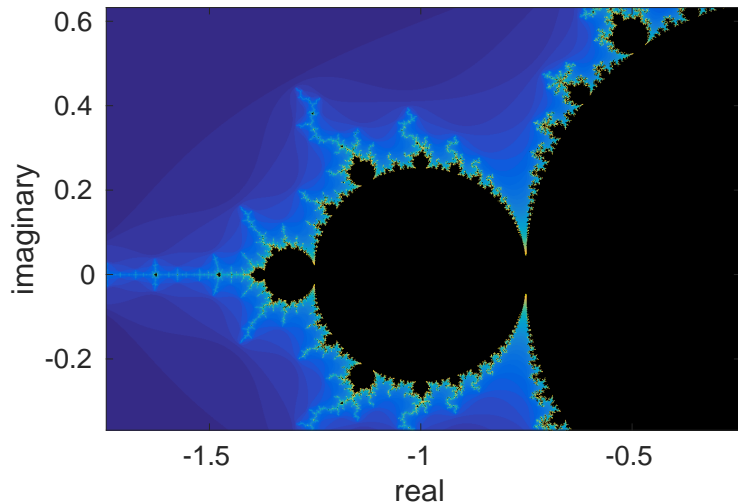
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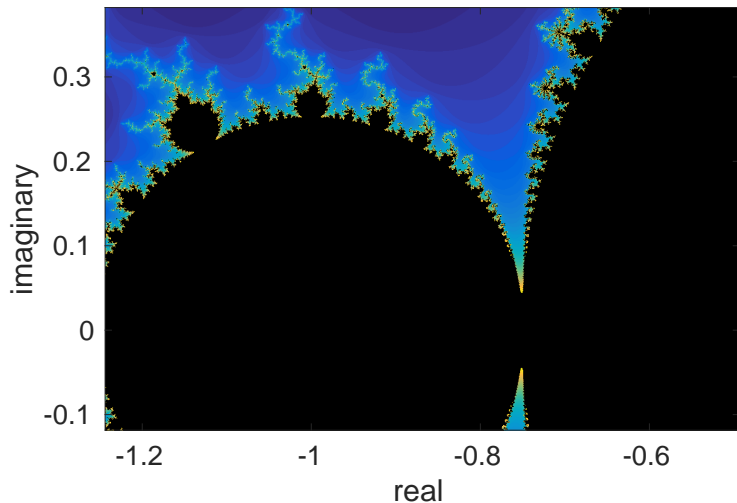
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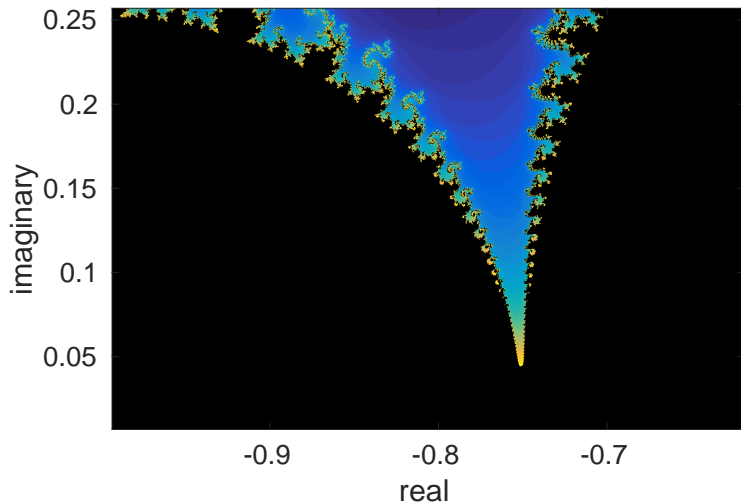
# Mandelbrot Set: zooming in



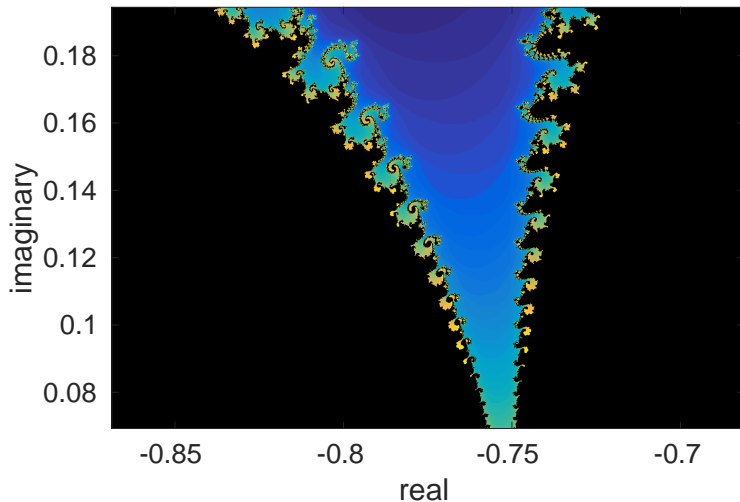
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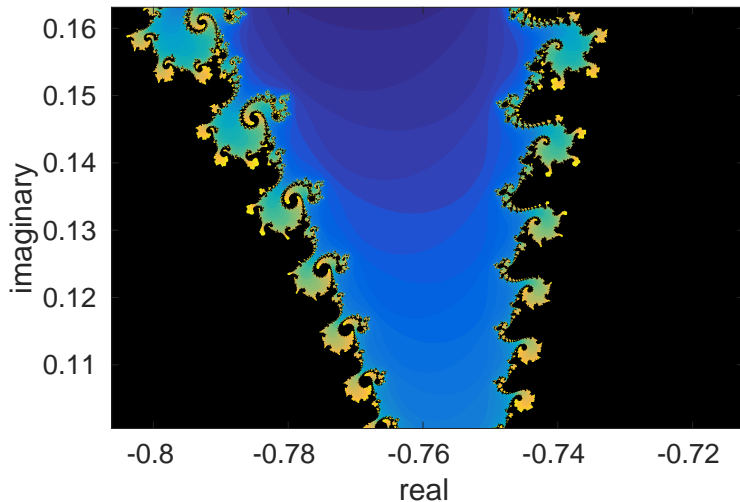
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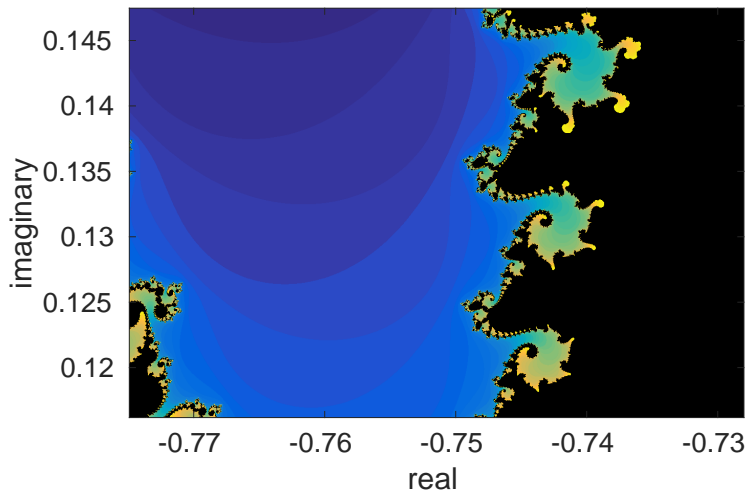
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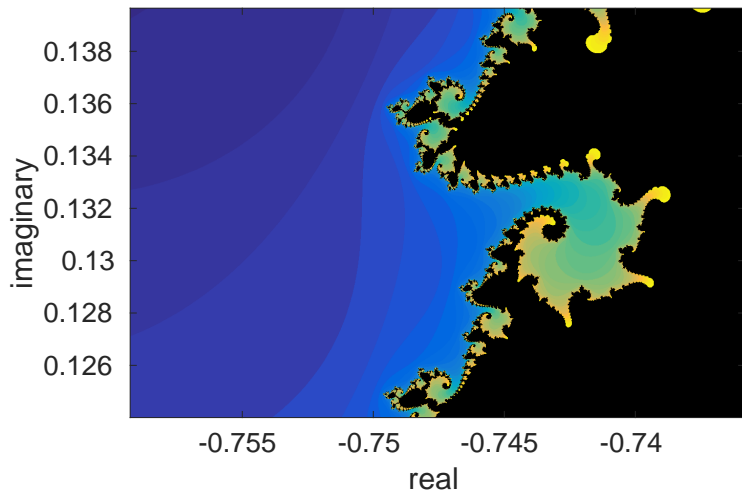


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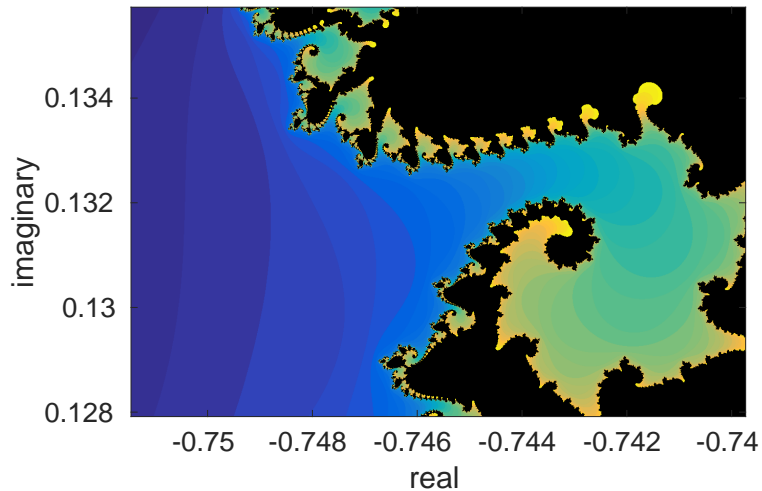




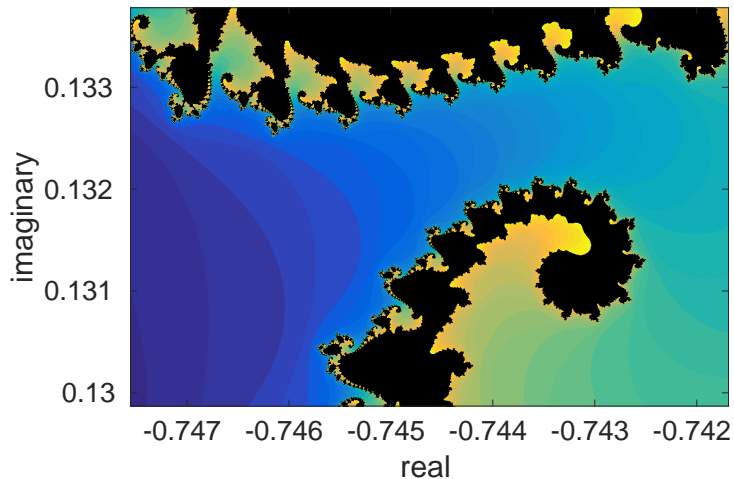
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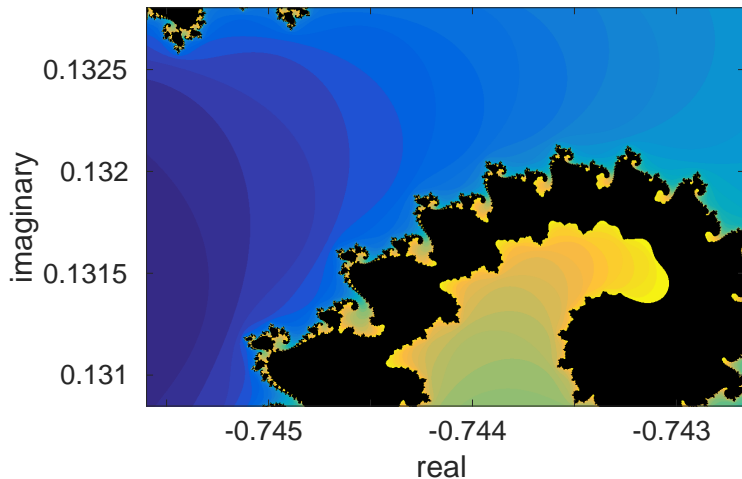
# Mandelbrot Set: zooming in



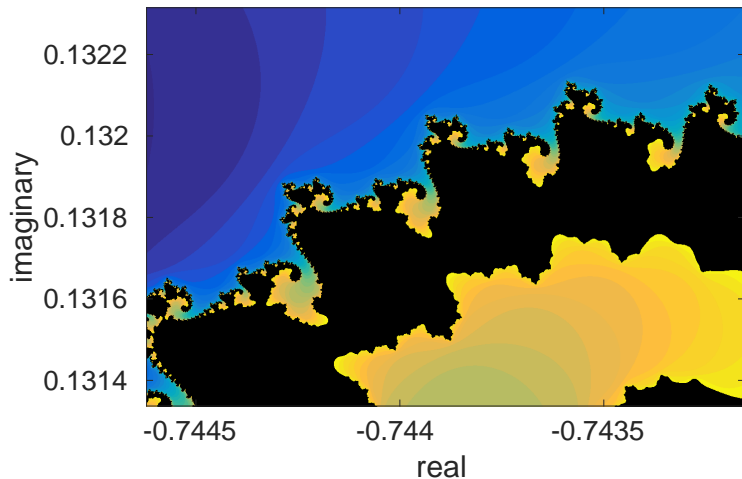
# Mandelbrot Set: zooming in



# Mandelbrot Set: zooming in



# Mandelbrot Set: zooming in



# Much more to learn

There is MUCH more to learn about fractals

- Fractional dimensions
- The relationship with *Chaos* and *strange attractors*
- Fractal compression
- Space filling curves
- L-systems
- More on IFSs

# Fractals and Art

I wonder whether fractal images are not touching the very structure of our brains. Is there a clue in the infinitely regressing character of such images that illuminates our perception of art? Could it be that a fractal image is of such extraordinary richness, that it is bound to resonate with our neuronal circuits and stimulate the pleasure I infer we all feel?

*P. W. Atkins*

- Balance between complexity and structure, predictability and randomness, is appealing to us
- Fractal patterns are aesthetically pleasing and stress-reducing  
Stress reduction from looking at fractals has been measured
- Fractal analysis has been used to separate fakes

# Fractals and Art

## “Accidental” examples

- Leonardo da Vinci's **Turbulence** (c1500)
- Hokusai's **Great Wave off Kanagawa** (1826)
- Jackson Pollock's **Blue Pole's** (1952)  
Sometimes called “fractal expressionism”
- Architecture
  - ▶ [https://en.wikipedia.org/wiki/Virupaksha\\_Temple,\\_Hampi](https://en.wikipedia.org/wiki/Virupaksha_Temple,_Hampi)
- Data-driven art
- And almost any picture of a natural scene ...



# Fractals and Art

Deliberate examples ([https://en.wikipedia.org/wiki/Fractal\\_art](https://en.wikipedia.org/wiki/Fractal_art))  
in Pictures

- <https://www.creativebloq.com/computer-arts/5-eye-popping-examples-fractal-art-71412376>
- <https://www.smashingmagazine.com/2008/10/50-phenomenal-fractal-art-pictures/>
- [https://en.wikipedia.org/wiki/The\\_Beauty\\_of\\_Fractals](https://en.wikipedia.org/wiki/The_Beauty_of_Fractals)
- Kerry Mitchell <http://www.kerrymitchellart.com/> [Mit]  
<https://www.fractalus.com/info/manifesto.htm> and  
[https://en.wikipedia.org/wiki/Kerry\\_Mitchell](https://en.wikipedia.org/wiki/Kerry_Mitchell)

# Fractals and Art

Deliberate examples ([https://en.wikipedia.org/wiki/Fractal\\_art](https://en.wikipedia.org/wiki/Fractal_art)) elsewhere

- Movies

- ▶ Fractals in Guardians of the Galaxy Vol 2.
- ▶ Dr Strange (also back to comics of Steve Ditko)

- Music

- ▶ Bruno Degazio <http://www-acad.sheridanc.on.ca/~degazio/AboutMeFolder/MusicPages/musiccomp.html> [Deg86]
- ▶ <https://quod.lib.umich.edu/s/spobooks/bbv9810.0001.001>

- Games

- ▶ Fractal's used in procedural generation software, e.g., Terragen <http://planet-side.co.uk/whats-new-in-terrigen-4/>
- ▶ more examples [https://en.wikipedia.org/wiki/Scenery\\_generator](https://en.wikipedia.org/wiki/Scenery_generator)

# Fractals in Inkscape

- Inkscape has several fractal-generation tools

- ▶ L-systems

- <http://people.cornellcollege.edu/dsherman/inkscape-fractal.html>

- <https://thebrickinthesky.wordpress.com/2013/03/17/l-systems-and-penrose-p3-in-inkscape/>

- <http://tavmjong.free.fr/INKSCAPE/MANUAL/html/Paths-LivePathEffects-VonKoch.html>

- ★ von Koch snowflakes
    - ★ Sierpinski gaskets
    - ★ ...

- ▶ Fractalize extension

- <http://prosepoetrycode.potterpcs.net/2015/08/fractal-rivers-with-inkscape/>

- More on this now.

## Section 2

# Statistical Self-Similarity

# Fractals and Modelling

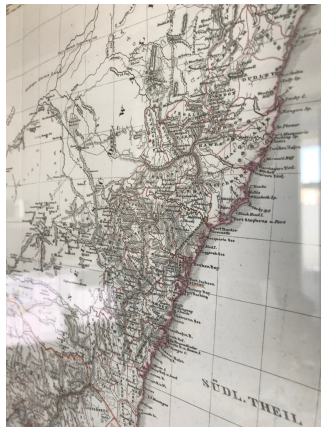
- Fractals (mostly) are deterministic
  - ▶ sometimes they look complex enough to be random
  - ▶ but the “model” is still deterministic
- Many observed phenomena have similar characteristics but are not at all regular
  - ▶ Geographic features: *e.g.*, mountains, rivers, coastlines
  - ▶ Biological systems: *e.g.*, blood vessels
  - ▶ Time series: rainfall patterns, cardiac rhythms, ...
  - ▶ And more: *e.g.*, clouds, snowflakes, ...

How do we model these?

# Statistical Self-Similarity Application: Coastline

Coastline paradox: the length of the coastline depends on the length of the rule you use to measure it [Ric61, Man67].

Known as the Richardson effect.

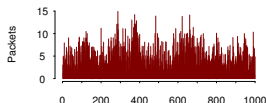
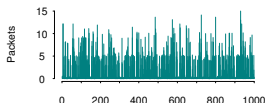


# Statistical Self-Similarity Application: Internet Traffic

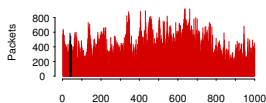
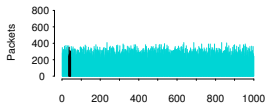
Traditional Model,  $H=0.5$

Real Data,  $H\sim 0.8$

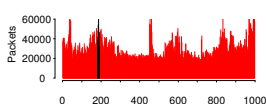
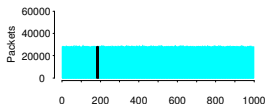
Time Unit = 0.01 Second



Time Unit = 1 Second



Time Unit = 100 Seconds



# Statistical Self-Similarity

A “statistic” of a “process” or *time series*  $X_t$  is a number that we derive as a summary or measurement of that process

- Mean or average

$$\mu_X = \mathbb{E}[X]$$

- Variance (similarly the standard deviation)

$$\sigma_X^2 = \text{Var}(X)$$

- Lots of others
  - ▶ autocovariance
  - ▶ spectrum (from Fourier transform)



# Statistical Self-Similarity

First, shift the time series so that it has mean 0, *i.e.*, subtract  $\mu_X$ .

- we often do this as a first step in analysis, because the long-term average is the first thing we look at, and then we analyse whatever was left.

Now imagine forming a *aggregated* time series  $\{X_k^{(m)}\}$  at level  $m$  by grouping the data into blocks of length  $m$ , and averaging

$$X_k^{(m)} := \frac{X_{(k-1)m+1} + \cdots + X_{km}}{m}.$$

Linearity of the expectation operator means that

$$\mathbb{E} \left[ X_k^{(m)} \right] = 0$$

But what about the other statistics?

# Statistical Self-Similarity

If, the process satisfies a scaling relationship such that

$$m^{1-H} X^{(m)}$$

has the same statistics (e.g., variance) as the original  $X^{(1)}$  then we say the process is *statistically self-similar* with Hurst parameter  $H$ .

# Statistical Self-Similarity Example

The  $1/f^\alpha$  noise we looked at before with spectrum

$$f_x(s) \sim c_f |s|^{-\alpha}, |s| \rightarrow 0$$

is statistically self-similar with Hurst parameter  $\alpha = 2H - 1$  or

$$H = \frac{\alpha + 1}{2}$$

- the case  $H = 0.5$  corresponds to the central limit case, or  $\alpha = 0$  or “white” noise
- but can have  $0 < H \leq 1$  as well; for  $H > 0.5$  we get “pink” noise, ...

## Mid-point displacement algorithm [FFC82]

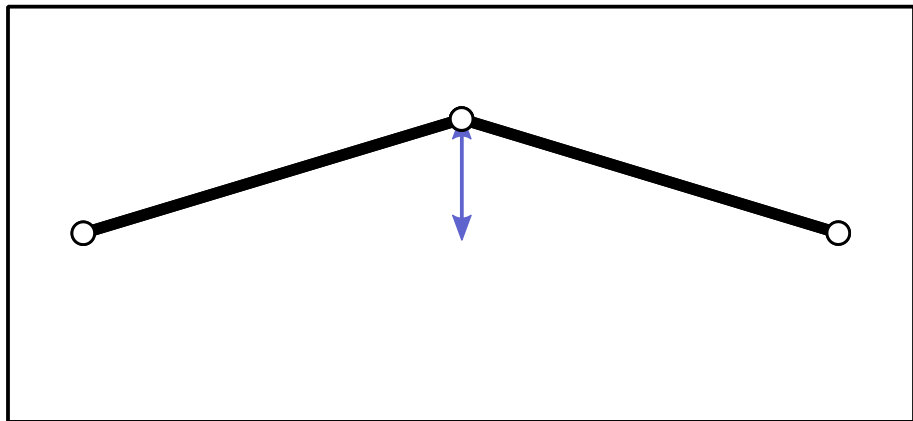
- Take  $H \in [1, 2]$ , and a variance  $\sigma$ , and define two end points.
- Displace the mid-point by a random amount with mean 0 and variance  $\sigma^2$
- Repeat this for each of the resulting line segments, but scale the variance by  $1/2^{2H}$
- Continue this process recursively

<http://old.cescg.org/CESCG97/marak/node3.html>

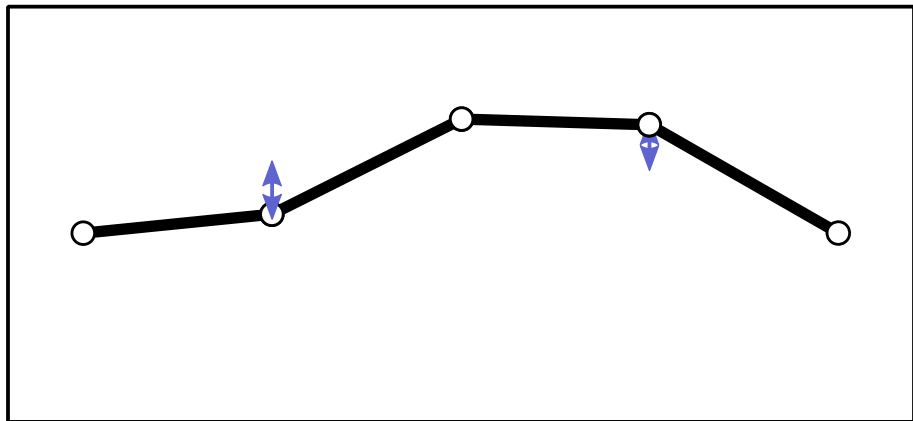
## Mid-point displacement algorithm example



## Mid-point displacement algorithm example

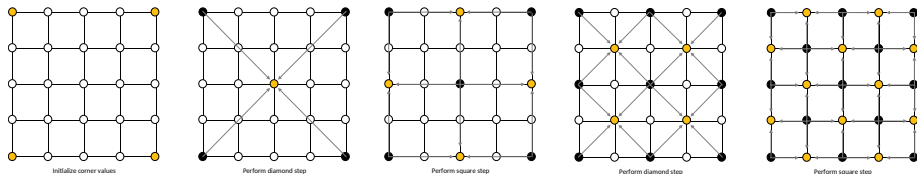


## Mid-point displacement algorithm example



# Mid-point displacement algorithm in 2D [FFC82]

Sometimes called the diamond-square algorithm



[https://en.wikipedia.org/wiki/Diamond-square\\_algorithm](https://en.wikipedia.org/wiki/Diamond-square_algorithm)

Used to generate “terrain”

[http:](http://planetside.co.uk/free-downloads/terragen-4-free-download/)

[//planetside.co.uk/free-downloads/terragen-4-free-download/](http://planetside.co.uk/free-downloads/terragen-4-free-download/)



# Mid-point displacement algorithm

## Variants

- Move location of mid-point in  $x$  and  $y$  directions
- Make boundaries periodic
- Higher dimensions (3D = clouds?)
- Approximate an underlying shape, with some smoothness
- Add colours
- Add dynamics (crumbled paper)

# Takeaways

- Fractals: complexity and non-locality from simple, local rules
- Statistical Self-Similarity: empirical modelling of real self-similar data
- Fractals are an empirical model, *i.e.*, they model phenomena we see, but often not an explanatory model, *i.e.*, they don't explain why we see what we see
  - ▶ but you can go deeper here, it's just beyond the scope of this course

# Section 3

## Extras

## Further reading I



Elisa Conversano and Laura L. Tedeschini, *Sierpinsky triangles in stone, on medieval floors in Rome*, Journal of Applied Mathematics **4** (2011), no. 4.



Bruno Degazio, *Musical aspects of fractal geometry*, International Computer Music Conference, 1986.



Alain Fournier, Don Fussell, and Loren Carpenter, *Computer rendering of stochastic models*, Communications of the ACM **25** (1982), no. 6, 371–384.



Benoit Mandelbrot, *How long is the coast of Britain? statistical self-similarity and fractional dimension*, Science **156** (1967), no. 3775, 636–638.



L. Kerry Mitchell, *Techniques for artistically rendering space-filling curves*.



Lewis F. Richardson, *The problem of contiguity: An appendix to statistics of deadly quarrels*, Yearbook of the Society for the Advancement of General Systems Theory (1961), 139187.