

CS 428: Fall 2009

Introduction to Computer Graphics

Procedural modeling

Procedural modeling

- Towards realism
 - Complexity = work (e.g. 2D/3D content creation)
 - Idea: put burden of *work* on computer for modeling relevant but nonspecific *detail*
 - Small specification → large range of detail/structure **amplification**
 - Examples
 - Mountains, trees, rivers, lightning, clouds, fire

Mountains



Clouds

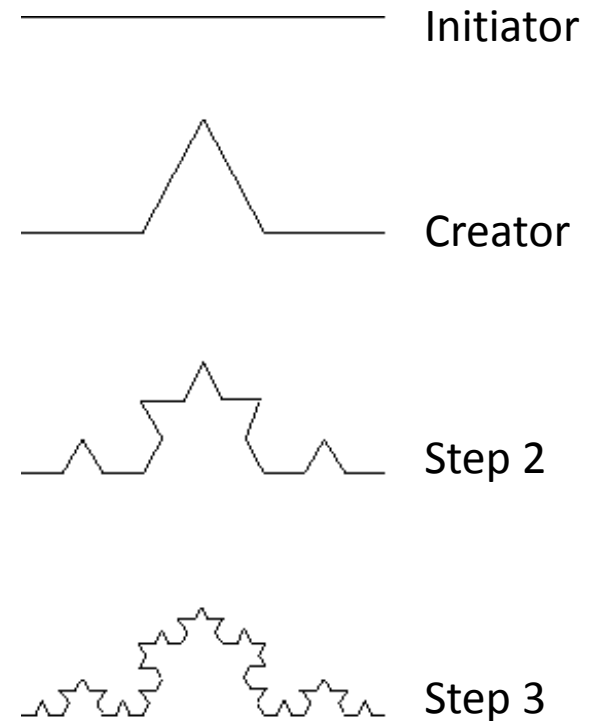


Fractals

- Common approach in CG
- *A language* for complexity of form
- An engineering approach
 - Modeling by structural similarity
 - not based on reality
- Definition
 - A geometrically complex object constructed via repetition over a range of scales (sizes):
leaves → trees → forests

Fractal self similarity

- Fractals have some geometrical scale invariance
- Example: Koch-curve
 - Each of the 4 line segments in the k-th step is a minified version of the entire curve in previous step by factor $1/3$
 - Cropped detail of the original curve can not be distinguished from the original



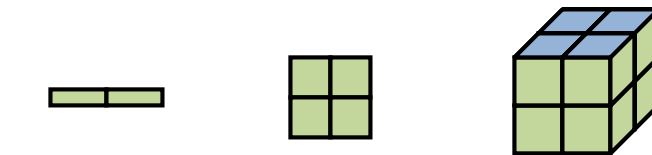
Fractal dimension

- For objects of dimensions 1, 2 and 3
- Subdivide into N equally sized parts

- Line $r = \frac{1}{N}$

- Square $r = \frac{1}{\sqrt{N}}$

- Cube $r = \frac{1}{\sqrt[3]{N}}$

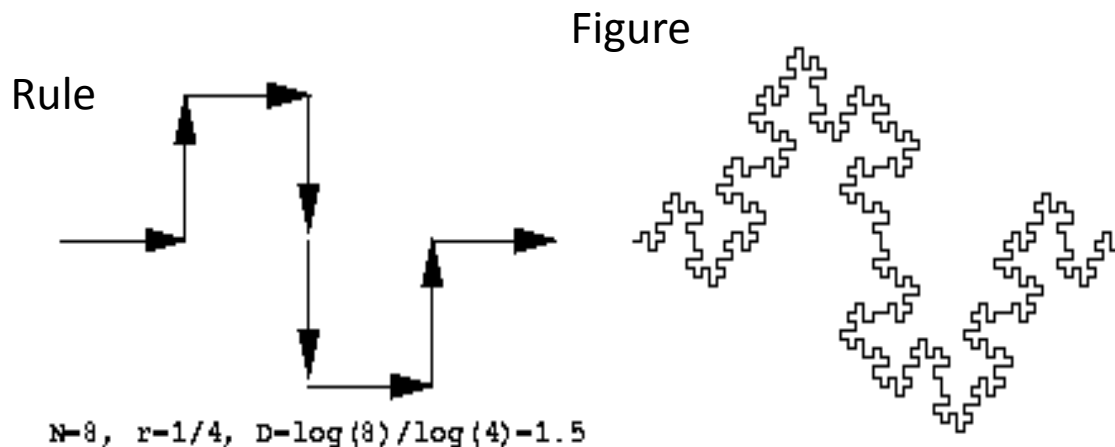


$$N = r^{-D} \iff D = \frac{\log N}{\log 1/r}$$

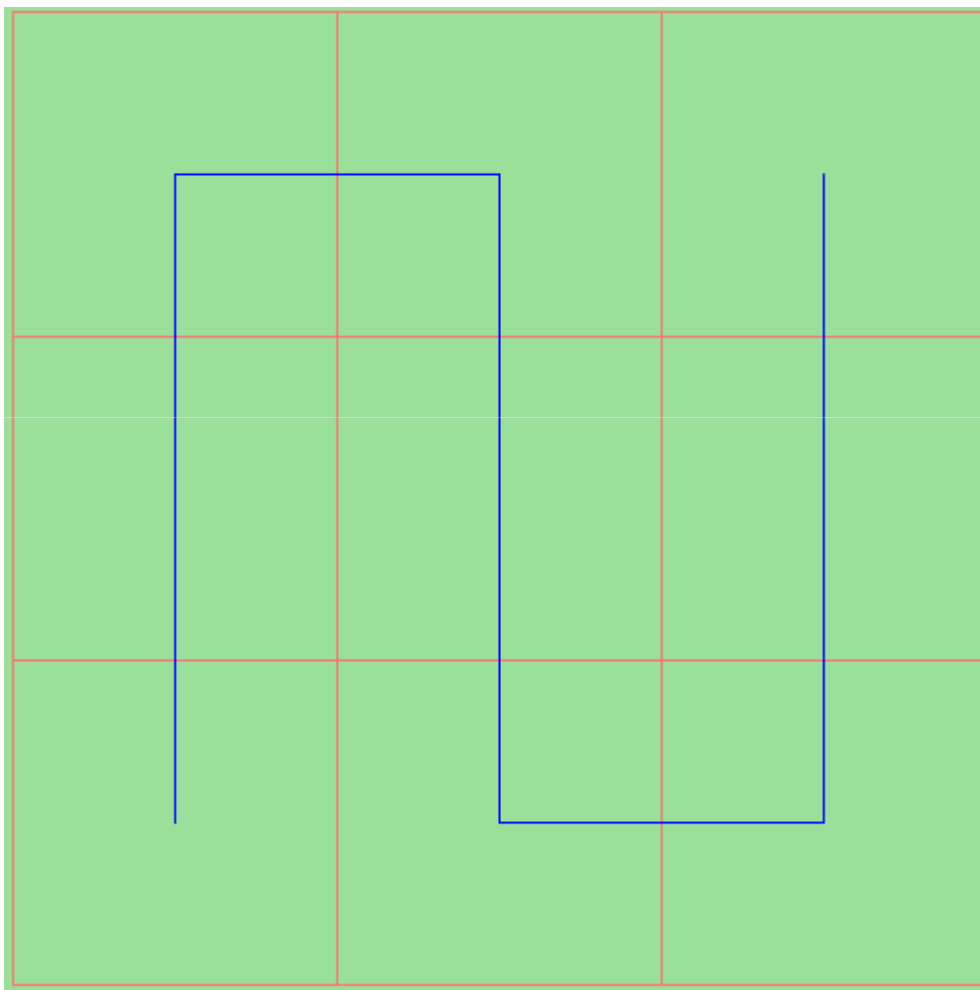
- A segment of the Koch-curve is made up from $N=4$ parts, each scaled by $r=1/3$ $D = \frac{\log 4}{\log 3} = 1.2619$

Fractal dimension

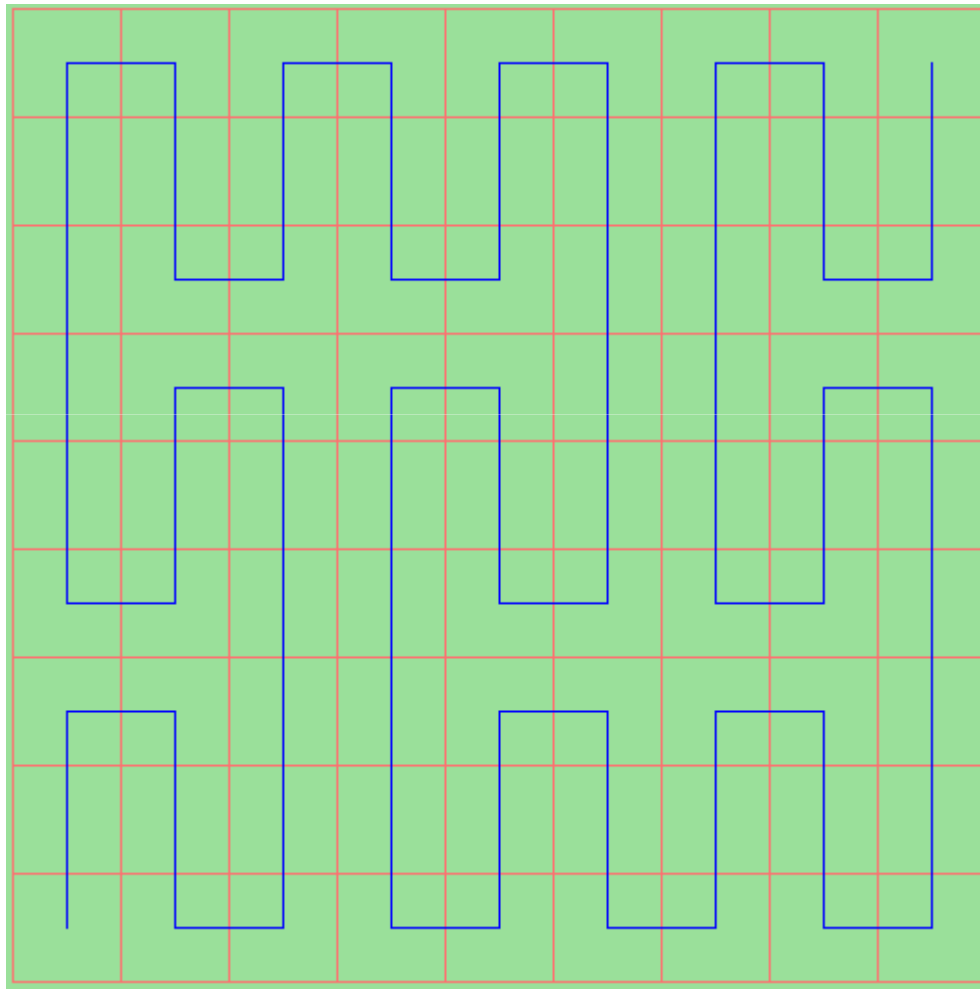
- Relation between number of parts N , and the associated scale factor r $D = \log N / \log 1/r$
- D is the fractal dimension or the self-similarity dimension of the structure (*roughness*)



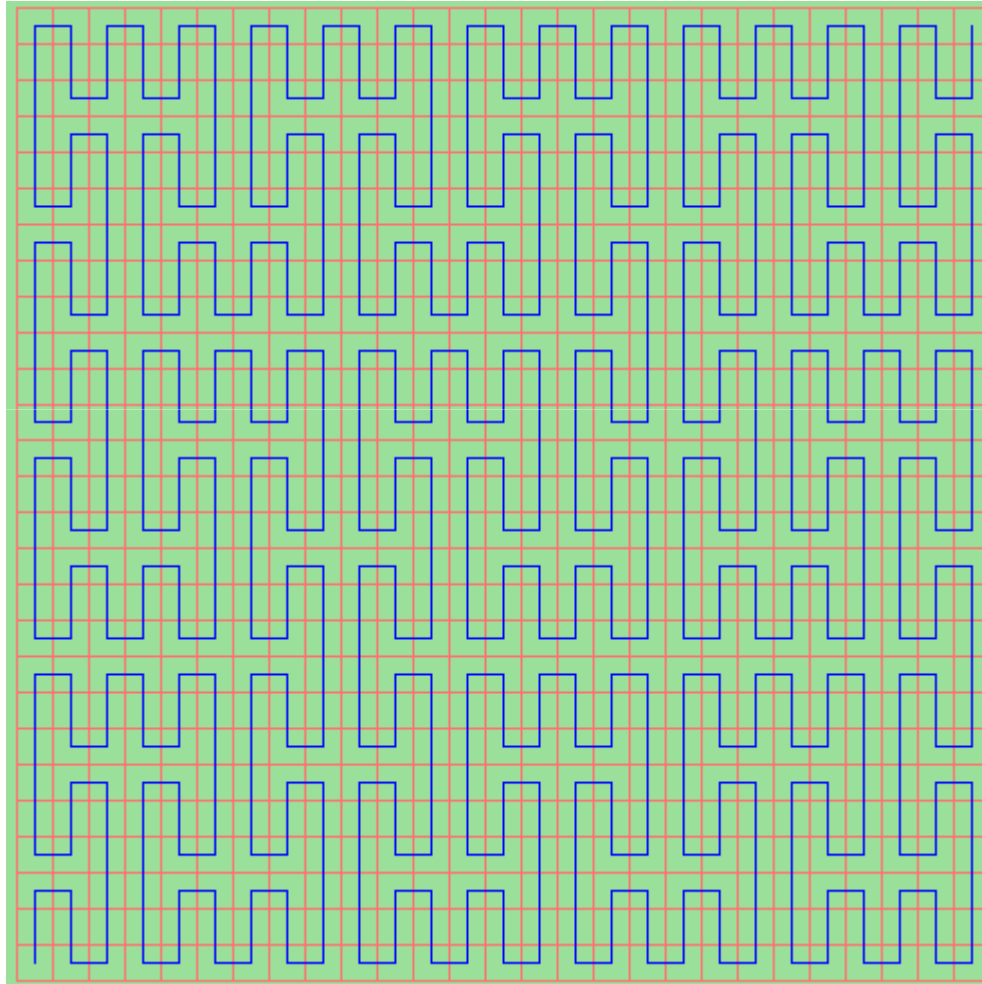
Peano space filling curves



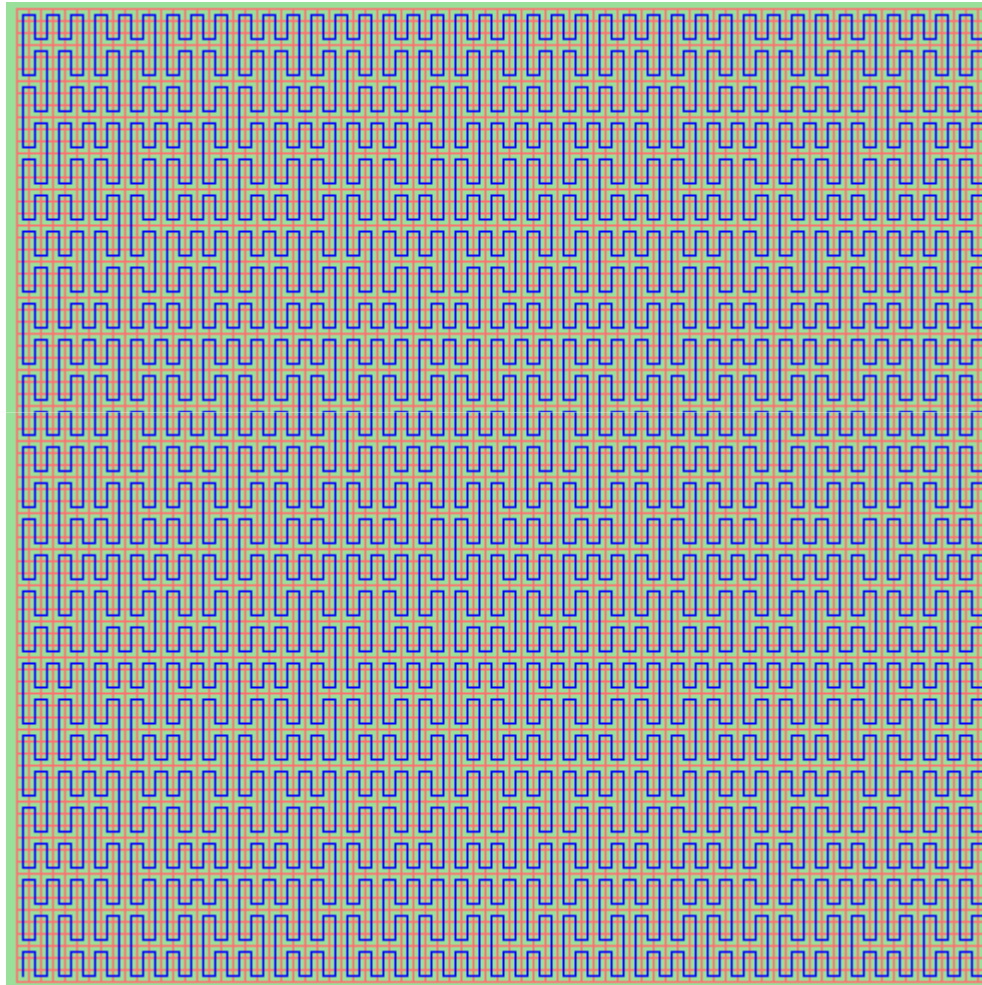
Peano space filling curves

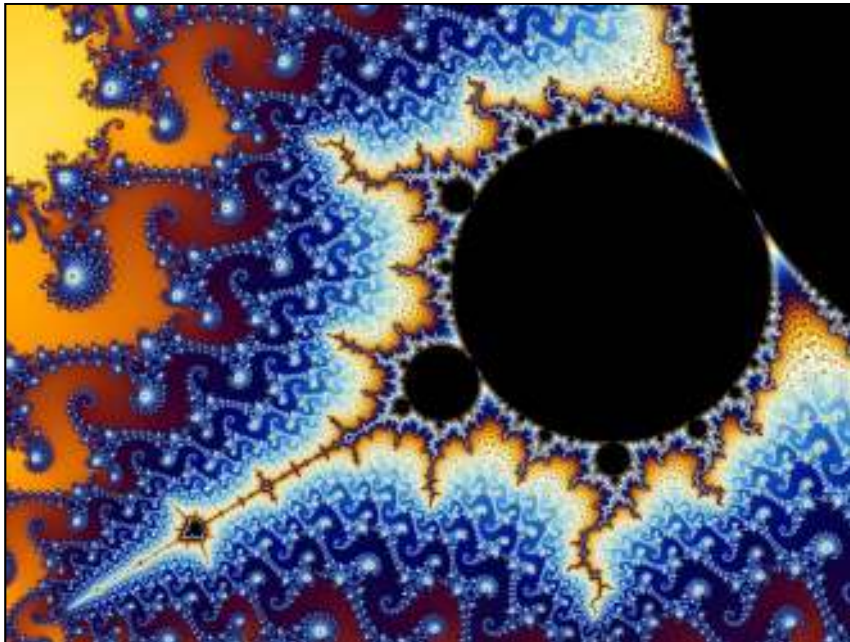
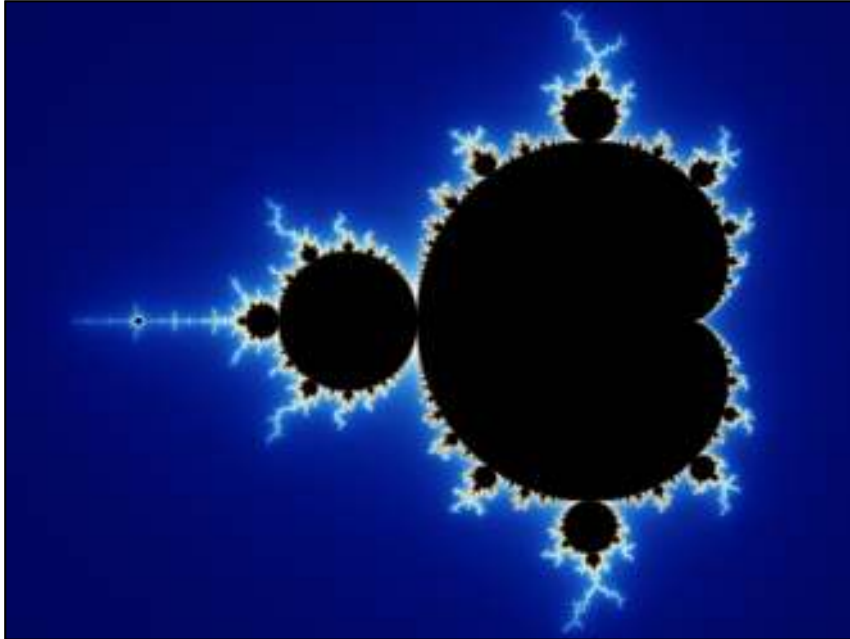


Peano space filling curves



Peano space filling curves





Mandelbrot set

```
float m( $x_0, y_0, i_{\max}$ )
```

```
   $x = x_0$ 
```

```
   $y = y_0$ 
```

```
  for  $i = 0$  to  $i_{\max}$ 
```

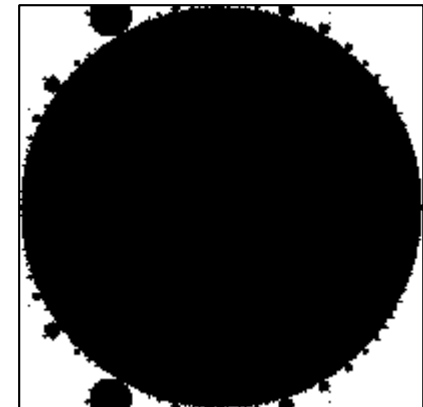
```
    if  $(x^2 + y^2 > 4)$  return  $i$ 
```

```
     $(x, y) \leftarrow (x_0 + x^2 - y^2, y_0 + 2xy)$ 
```

```
  end for
```

```
  return 0
```

```
end
```



Fractals

- Build complexity from repetition
- Structure repetition
 - Frequency, amplitude and lacunarity (space filling)
- Beneficial features
 - Fine structure at **all** scales
 - Not regular
 - Self similar
 - **Compact description**

Stochastistic self similarity

- Natural objects such as farns or coastlines are not exactly identical when magnified
 - But characteristics are similar
 - Magnified coast line similar to original
 - This phenomenon is captured by stochastic self-similarity



Stochastistic fractals

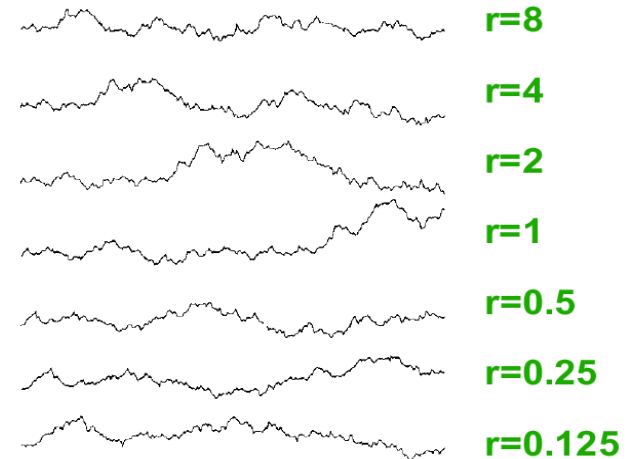
- Used to model
 - Terrain, clouds, waves, tree bark, etc.
 - Useful for generating 3D wood/marble textures
 - Simulation of **Brownian motion**



Brownian motion

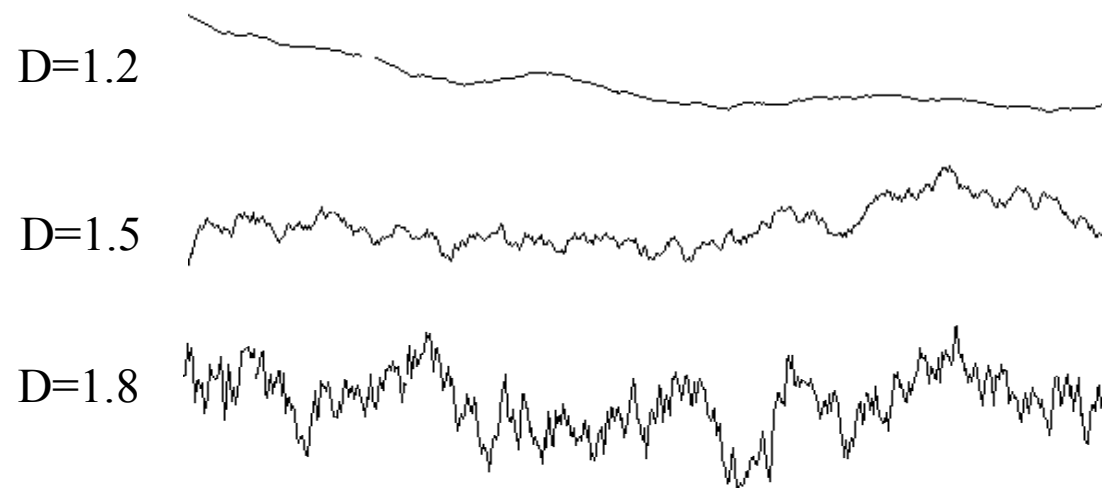
- Non-stationary stochastic process $\{X(t)\}$
- Increments $X(t+s)-X(t)$ are Gaussian distributed with expected value of zero
- Variance of increments is proportional to time difference $\rightarrow \text{var}(X(t+s)-X(t)) \sim |s|$
- Statistically self-similar

$X(t)$ Statistically similar to $\frac{1}{\sqrt{r}} X(rt)$



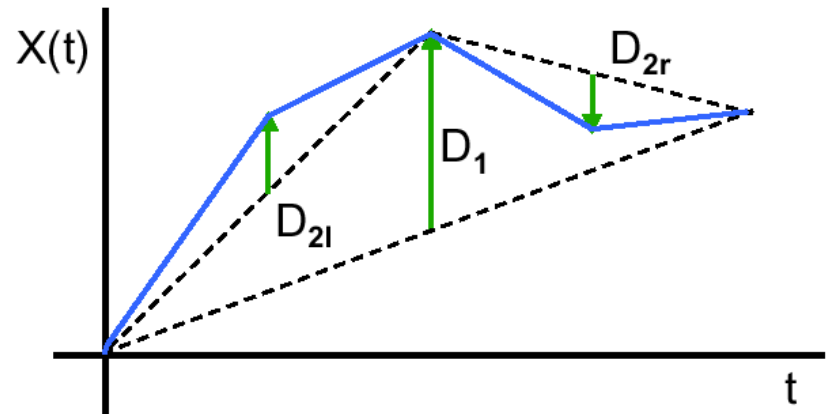
Fractal Brownian motion

- Brownian motion has fractal dimension 1.5
- Dimension D of natural objects ranges from 1.15 and 1.25
 - Use of fractal Brownian motion (fBm)



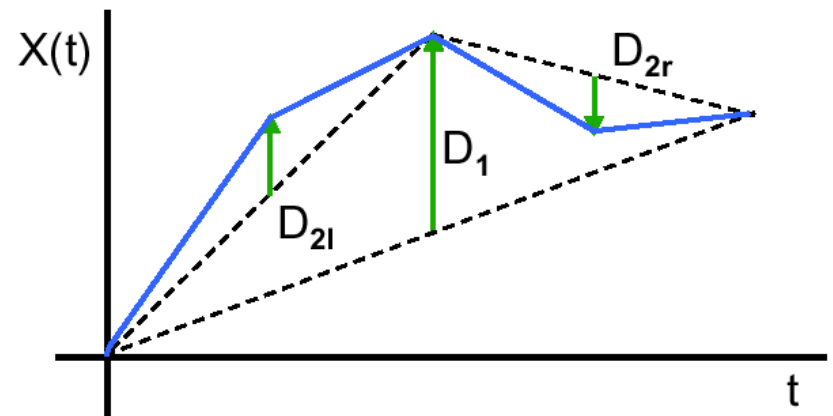
Brownian motion application

- Midpoint displacement (for $D \in [1,2]$)
 - Step 0: $X(1/2) = (X(0) + X(1))/2 + D_0$
 - Step n: linear interpolation between neighboring points and displacement by D_n
 - D_n is a Gaussian distributed random variable with $E(D_n)=0$ and $\text{var}(D_n)=(1-2^{2-2D})/2^{n(4-2D)}$
 - $X(t)$ and $1/2^{(2-D)}X(2t)$ are statistically self-similar



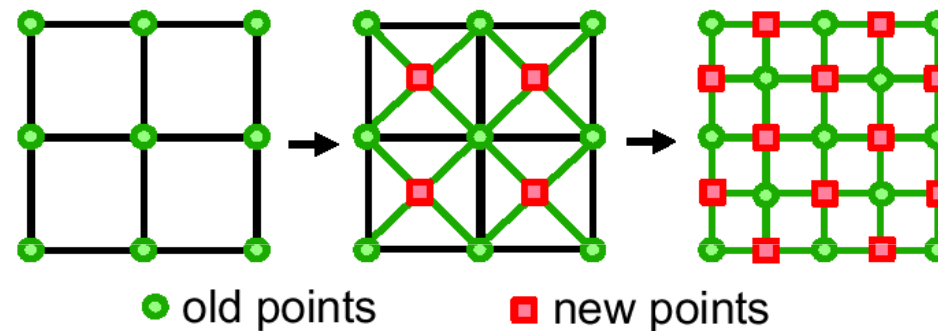
Brownian motion application

- Midpoint displacement (for $D \in [1,2]$)
- Case of $D=1.5$, with $E(D_n) = 0$ (mean = 0)
 - Start with $\text{var}(D_n)=0.5$ (for initial grid spacing of 1)
 - In each step, scale $\text{var}(D_n)$ by $1/2$
 - Same as $\text{var}(D_n)=0.5 \cdot (1/2)^n$
 - Use **Random** class in Java
 - `nextDouble()` (in $[0,1]$)
 - `nextGaussian()` (mean = 0, var = 1)



Midpoint displacement in 2D

- Create refined grid

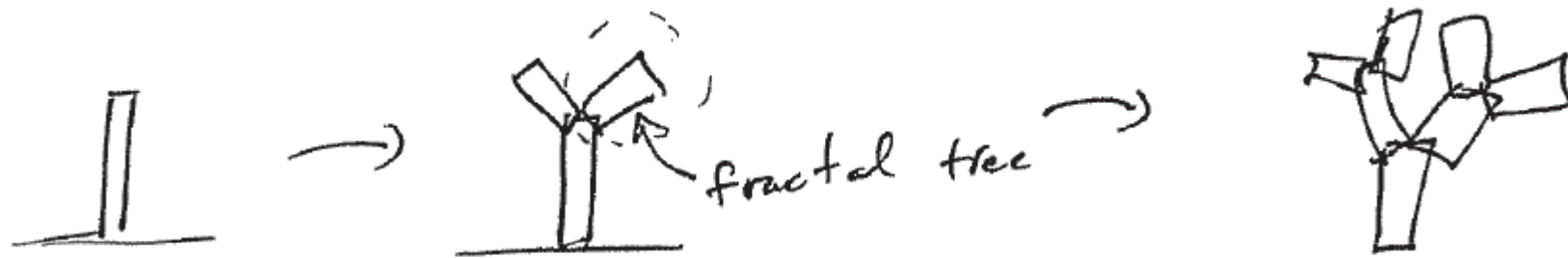


- In every step the grid resolution is scaled by a factor of $r = \frac{1}{\sqrt{2}}$
- The variance of random displacements is scaled by

$$r^{4-2D} = \left(\frac{1}{2}\right)^{2-D} \quad \text{for } D = 1.5 \quad r = \left(\frac{1}{2}\right)^{0.5} = \frac{1}{\sqrt{2}}$$

Fractal trees

- Hierarchical fractal modeling

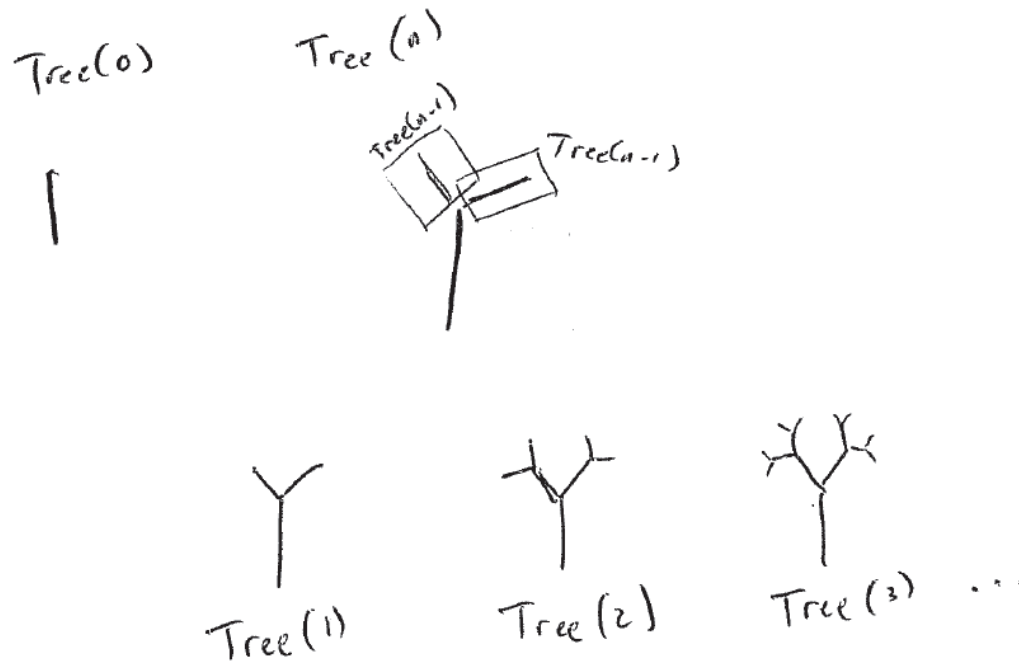
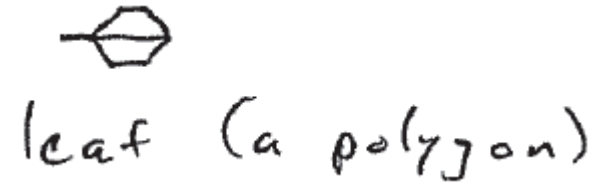


- Random angles, lengths, placements

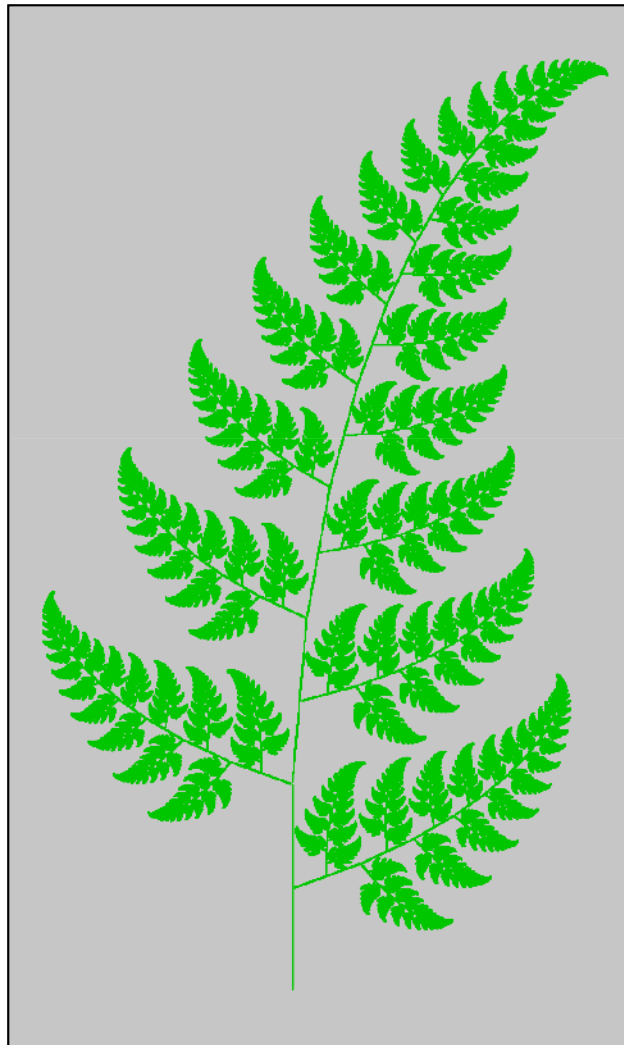


Fractal trees

- At end of recursion, add leaf
- Decrease branch length and radius as recursion depth increases



Fractal fern





Procedural textures

Perlin Noise

- Sum of band limited functions (octaves)
 - In uv space compute a pseudorandom number and a gradient
 - Interpolation (linear, cubic) between integer uv's provides a smooth band limited function

$$\frac{1}{6} * (\text{img}_1 + \text{img}_2 + \text{img}_3 + \text{img}_4 + \text{img}_5 + \text{img}_6) = \text{img}_7$$

The diagram illustrates the summation of six band-limited functions (octaves) to produce a final Perlin noise texture. The first row shows six square images of increasing frequency, each representing an octave. These are summed together, as indicated by the plus signs and the large parentheses. The result is shown in the second row as a single square image, which is the final Perlin noise texture. The scaling factor $1/6$ is applied to the entire sum.

Procedural textures

Perlin Noise

