

Scale Invariance and Universality

Chapter 5

Wednesday 8 May

Before we start:

This week and next week:

L1

Guest lecture Fernando

L2

Chap 5. L1: RW

Chap4. L2: Epidemic modeling

Chap 5. L2: renormalization + guest lecture

Wout Merbis,

from the Dutch Institute for Emergent Phenomena

Vítor Vasconcelos, from the Computational Science Lab

Quiz: Tuesday May 21 at 11h

Questions:

Comment: Bonus homework?

Information about the last quiz.

Issue about schedule of the exam? Except for *Machine learning for physicists*

Scale Invariance and Universality

Chapter 5

Plan: **Lecture 1:** Emergent behaviors in the simple example of random walks:
scale-invariance and universality

Lecture 2: Renormalization (next Thursday)

Tutorial: Random walks + (optional) renormalization

Emergent behaviors in the simple example of random walks: scale-invariance and universality

Chapter 5. L1

Plan: 1) Introduction: properties of critical phenomena

2) Fractals: notion of self-similarity and fractal dimension

3) Random walk and Scale invariance

Expectations: Participate in the discussions, take notes, try to do the analytical derivations

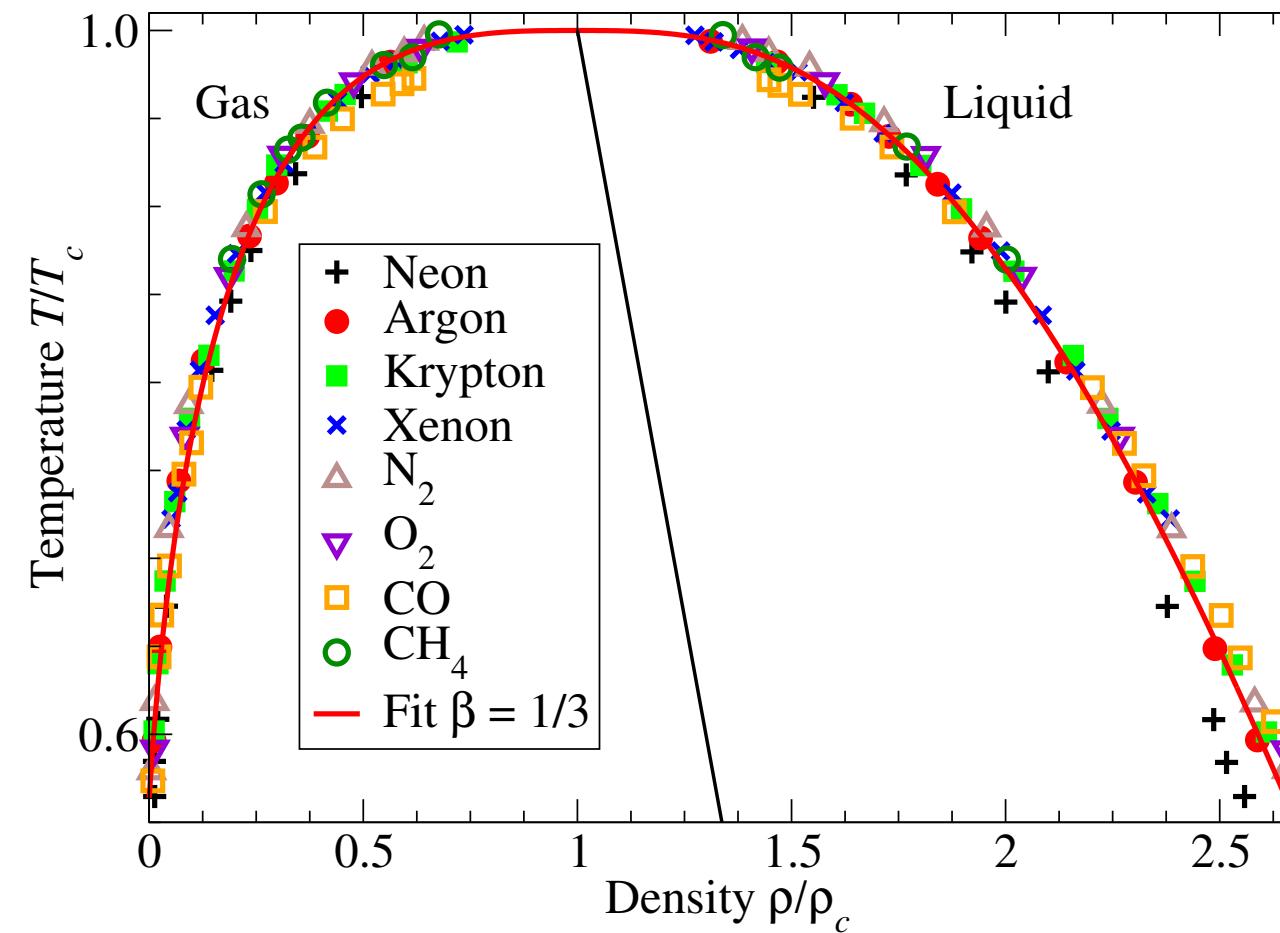
References: “[Entropy, Order Parameters, and Complexity](#)”, by J. P. Sethna, Chap. 2 and 12.2

Introduction

Do you remember main **properties** of **continuous phase transitions at criticality?**

Properties of Critical phenomena:

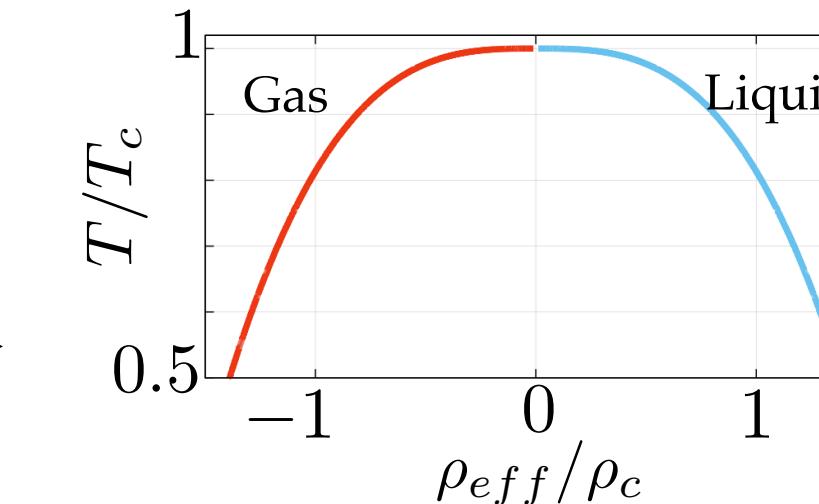
?? _____ ?? : two systems, microscopically completely **different**, can exhibit precisely the **same critical behavior near their critical point**.



Ex. _____ at the liquid-gas critical point:

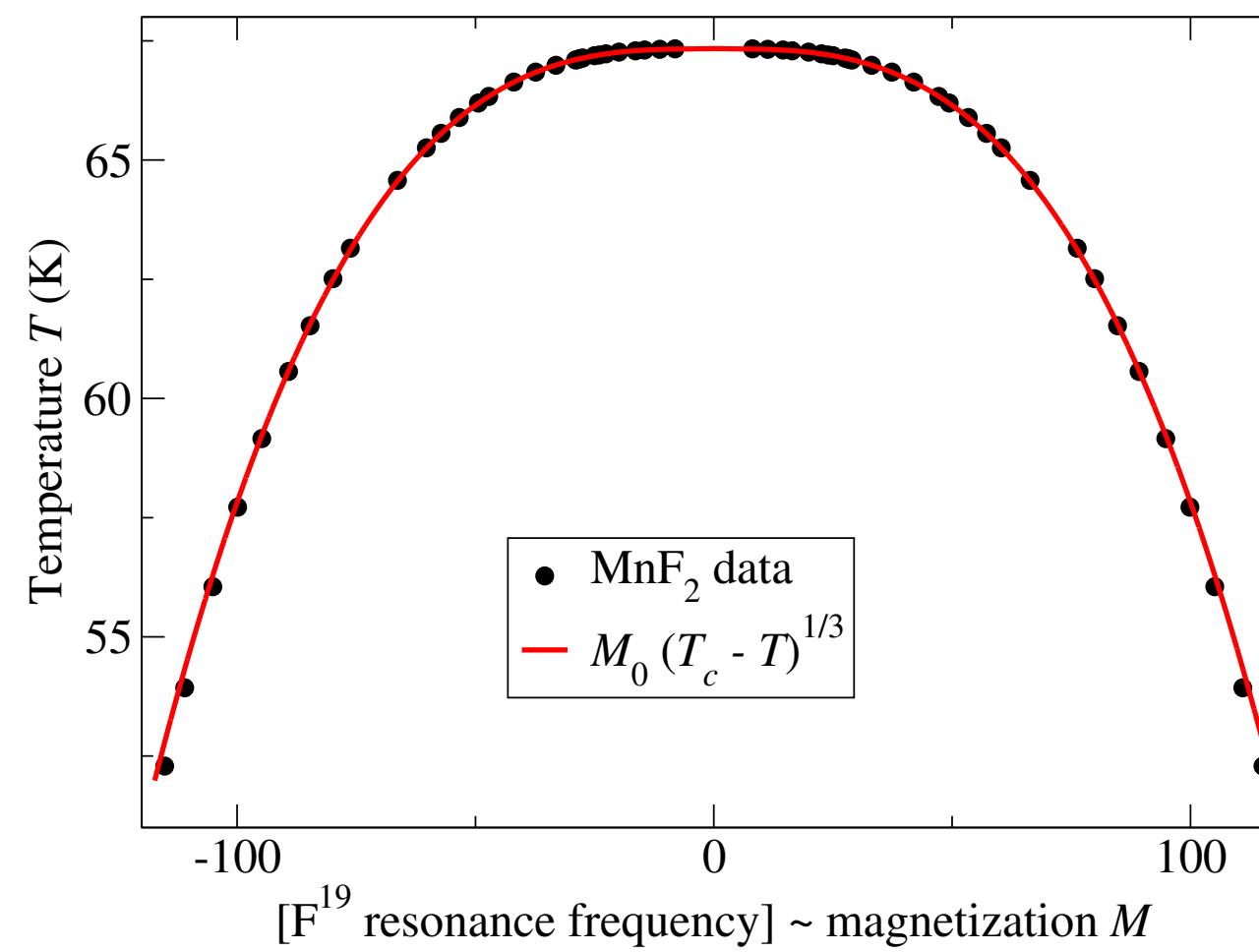
Liquid-gas coexistence lines for a variety of atoms and small molecules near their critical point.

Tilted



$$\frac{\rho_{eff}}{\rho_c} = \pm \rho_0 \left(1 - \frac{T}{T_c}\right)^\beta$$

$$\begin{cases} \rho_0 = 1.75 \\ \beta = 1/3 \end{cases}$$



Ex. _____ : 3D ferromagnetic-paramagnetic critical point

Uniaxial antiferromagnet MnF₂

$$\langle M \rangle = M_0 T_c \left(1 - \frac{T}{T_c}\right)^\beta$$

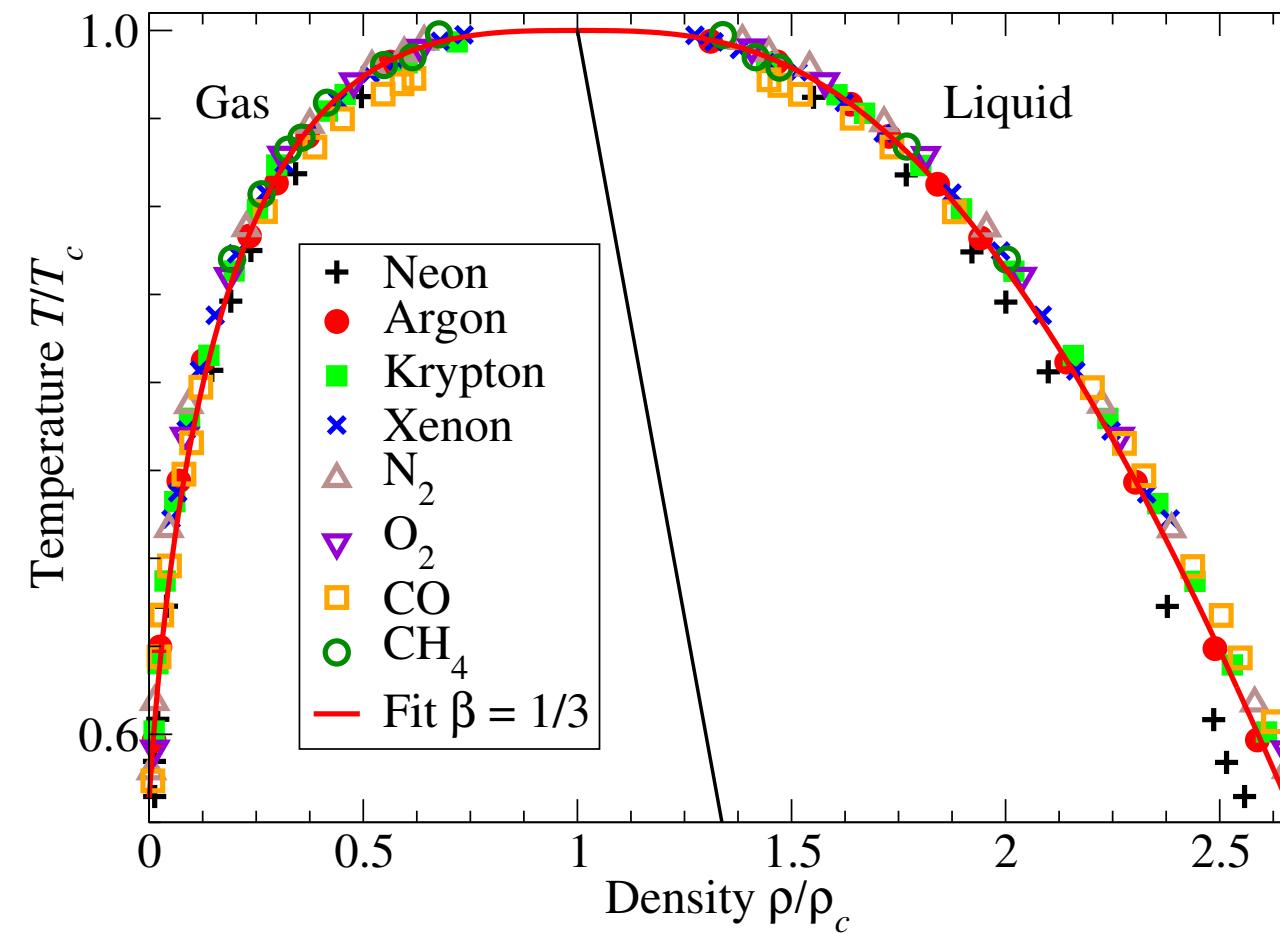
$$\beta = 1/3$$

Current estimate: $\beta = 0.3264\dots$

Critical exponents all identical: at the critical point, correlation length, susceptibility, specific heat have power-law singularities with the same exponents.

Properties of Critical phenomena: Universality!

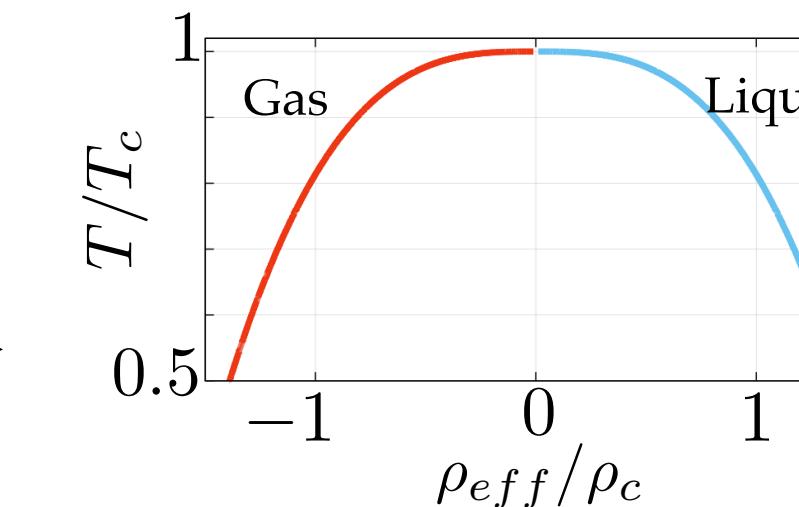
Universality: two systems, **microscopically completely different**, can exhibit precisely the **same critical behavior near their critical point**.



Ex. Universality at the liquid-gas critical point:

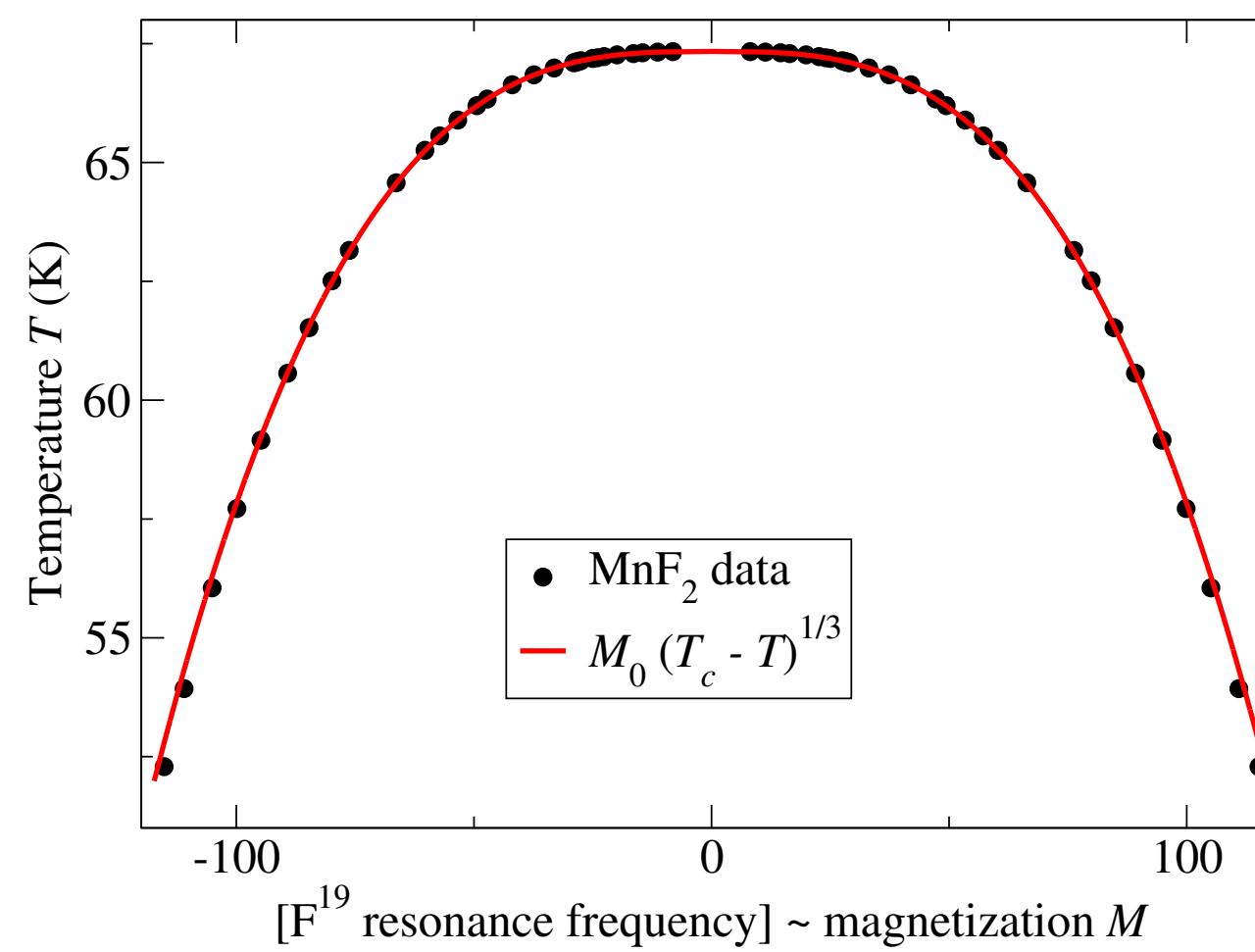
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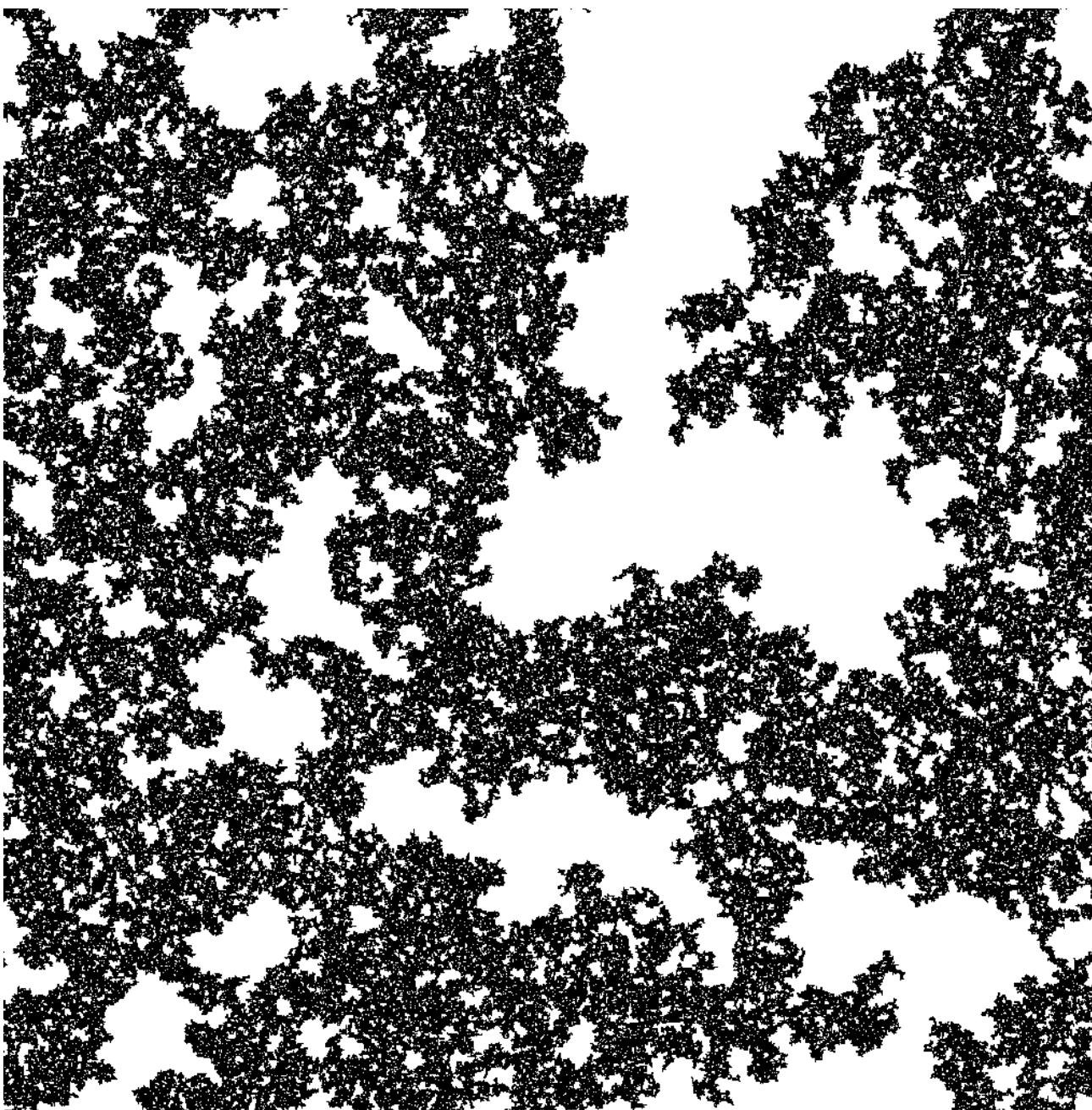
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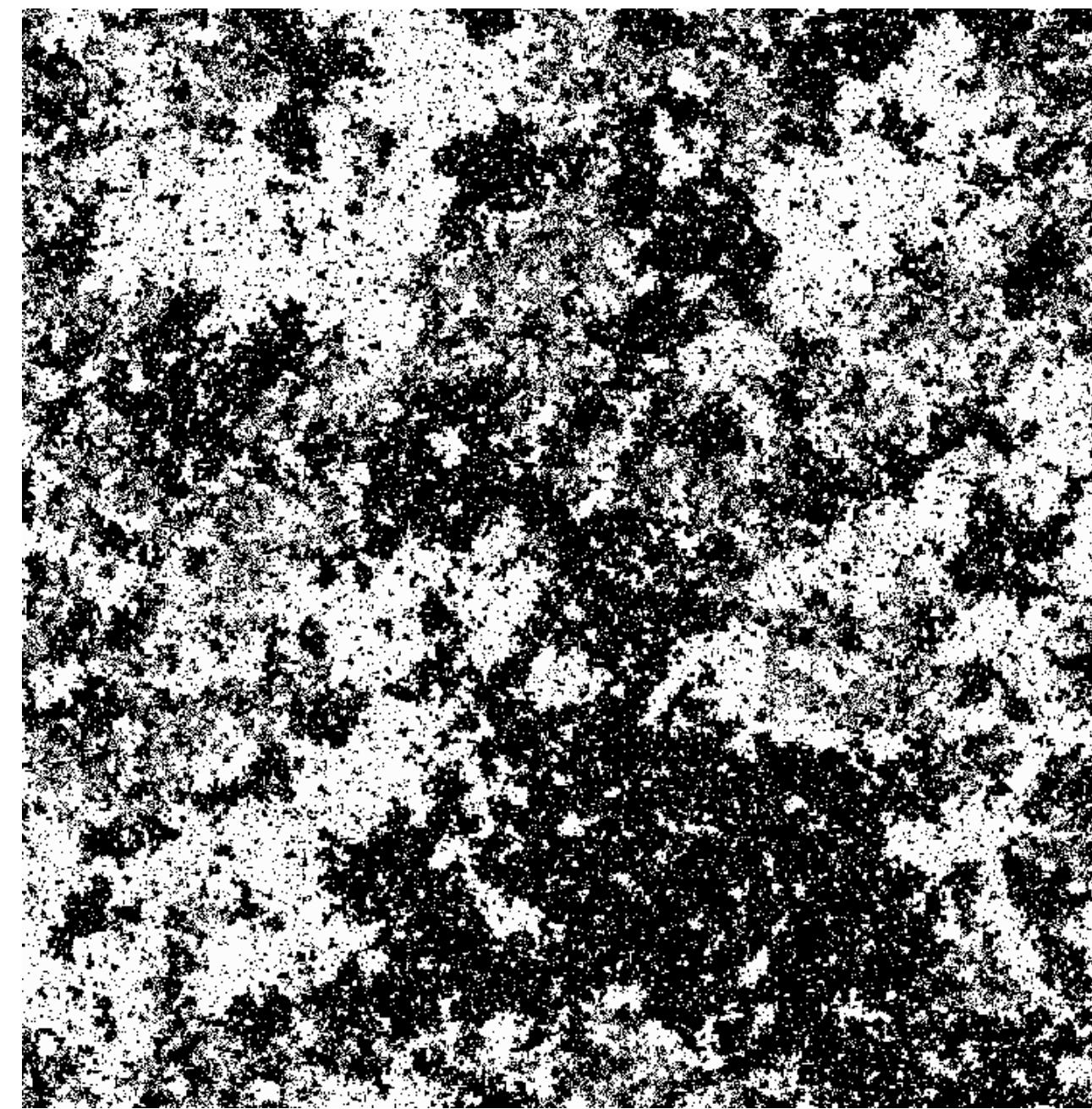
They are in the **same universality class**, along with the **3D Ising model**

(despite drastic simplifications!)

Properties of Critical phenomena: _____?



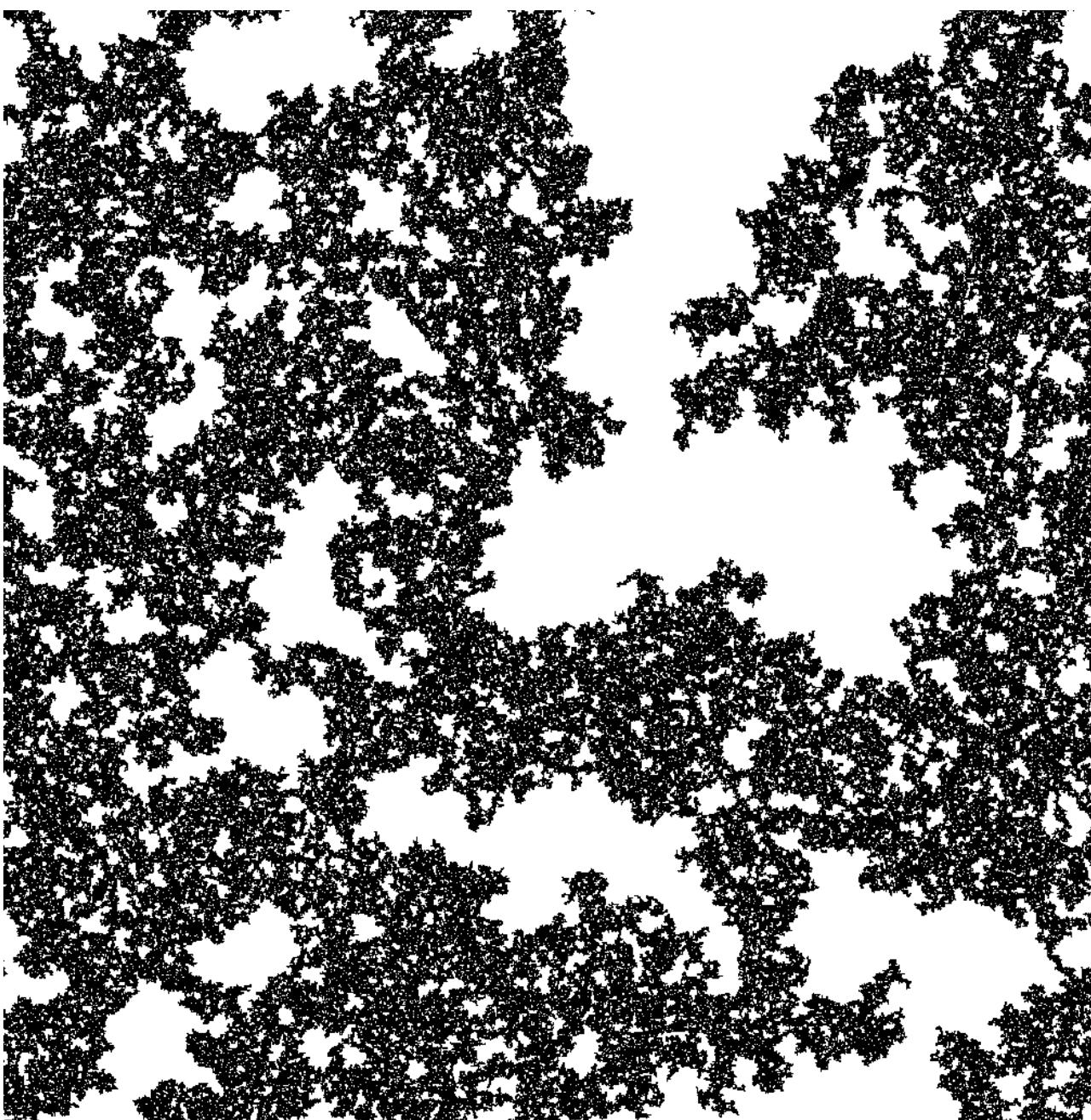
Bond percolation
at $p = p_c = 0.5$



Ising model
at $T = T_c \sim 2.269$

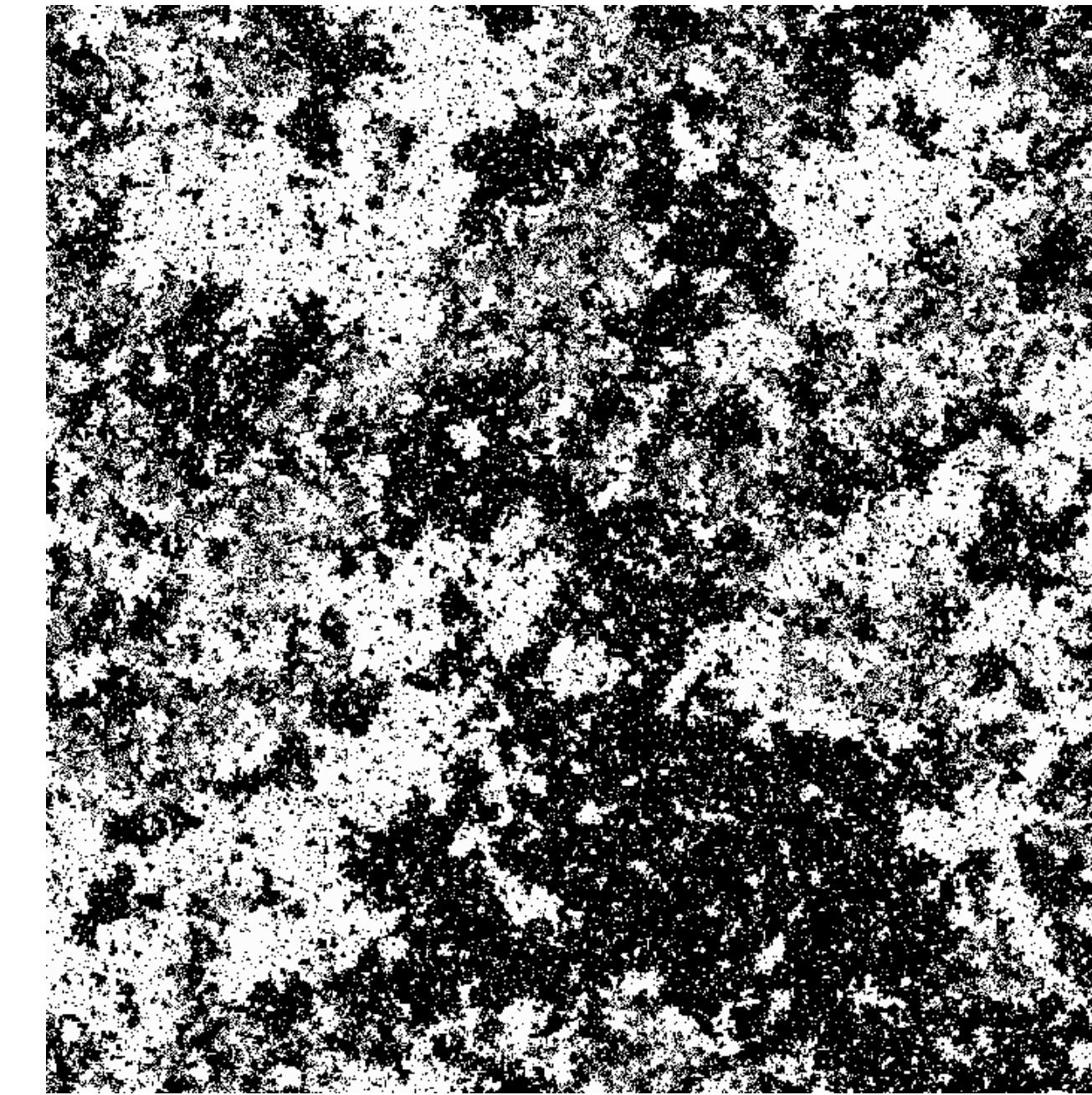
Anything interesting on these pictures?

Properties of Critical phenomena: _____?



Bond percolation

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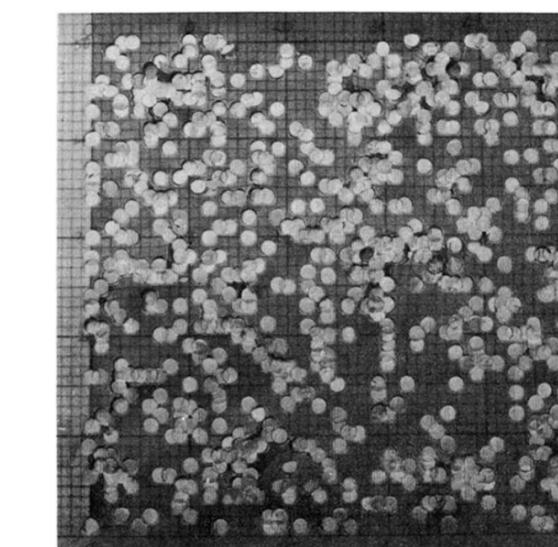


Ising model

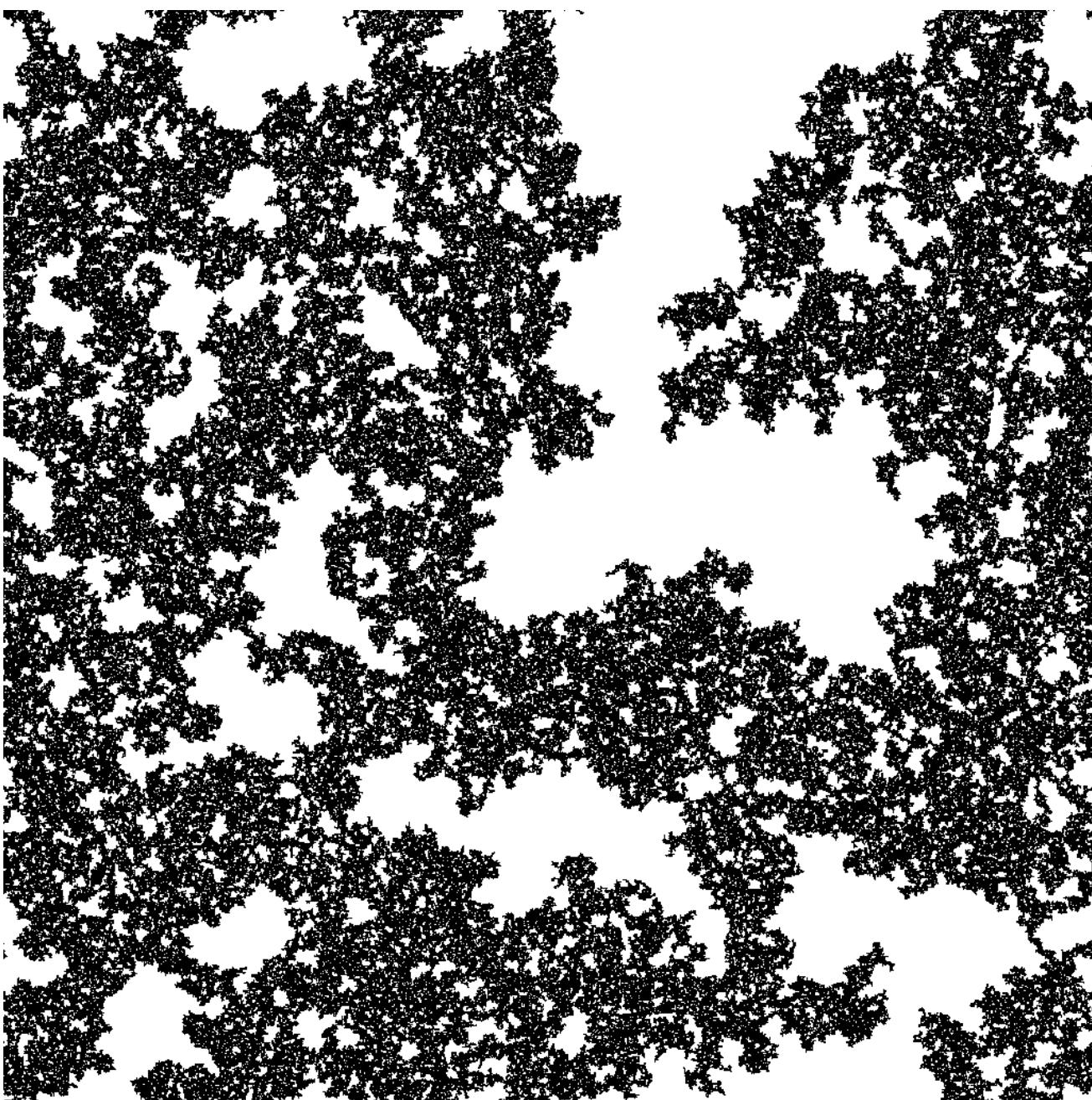
at $T = T_c \sim 2.269$

“Fractal-like” structure: very “tortuous” paths

cf. recall:
conductivity fall to small value close to critical concentration of wholes,
paths become very few and tortuous just before the sheet fall apart.

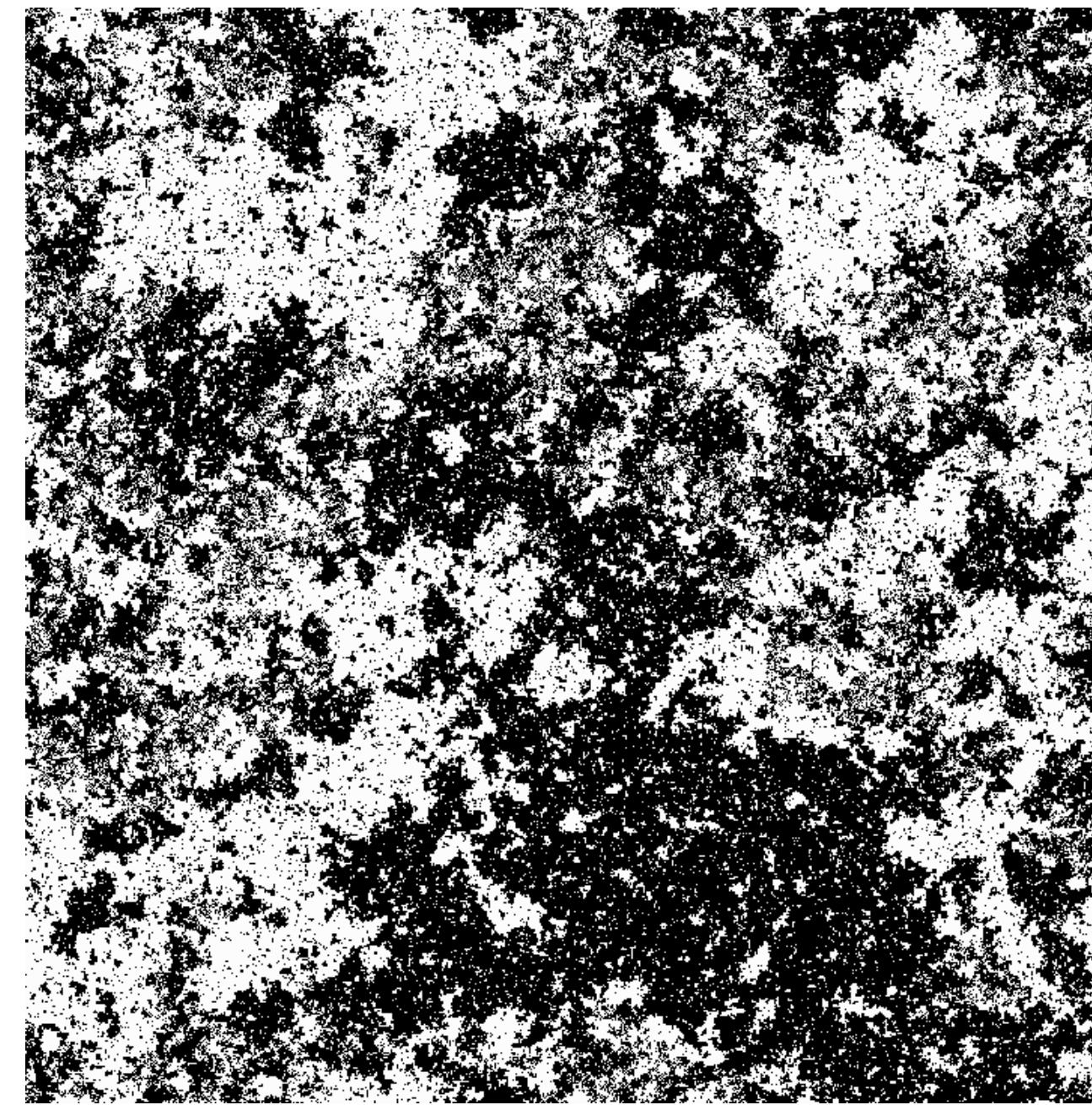


Properties of Critical phenomena: Scale Invariance



Bond percolation

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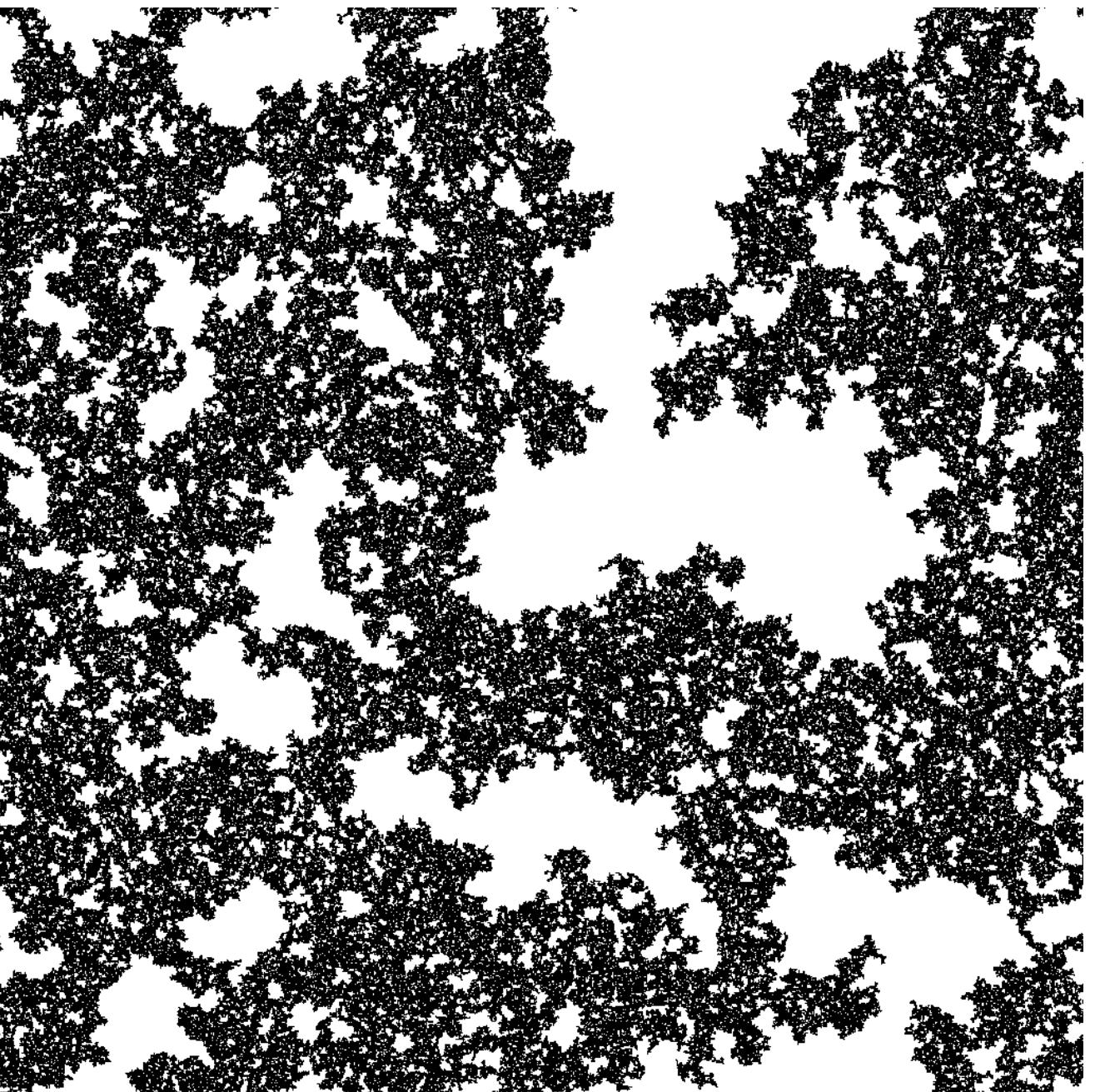


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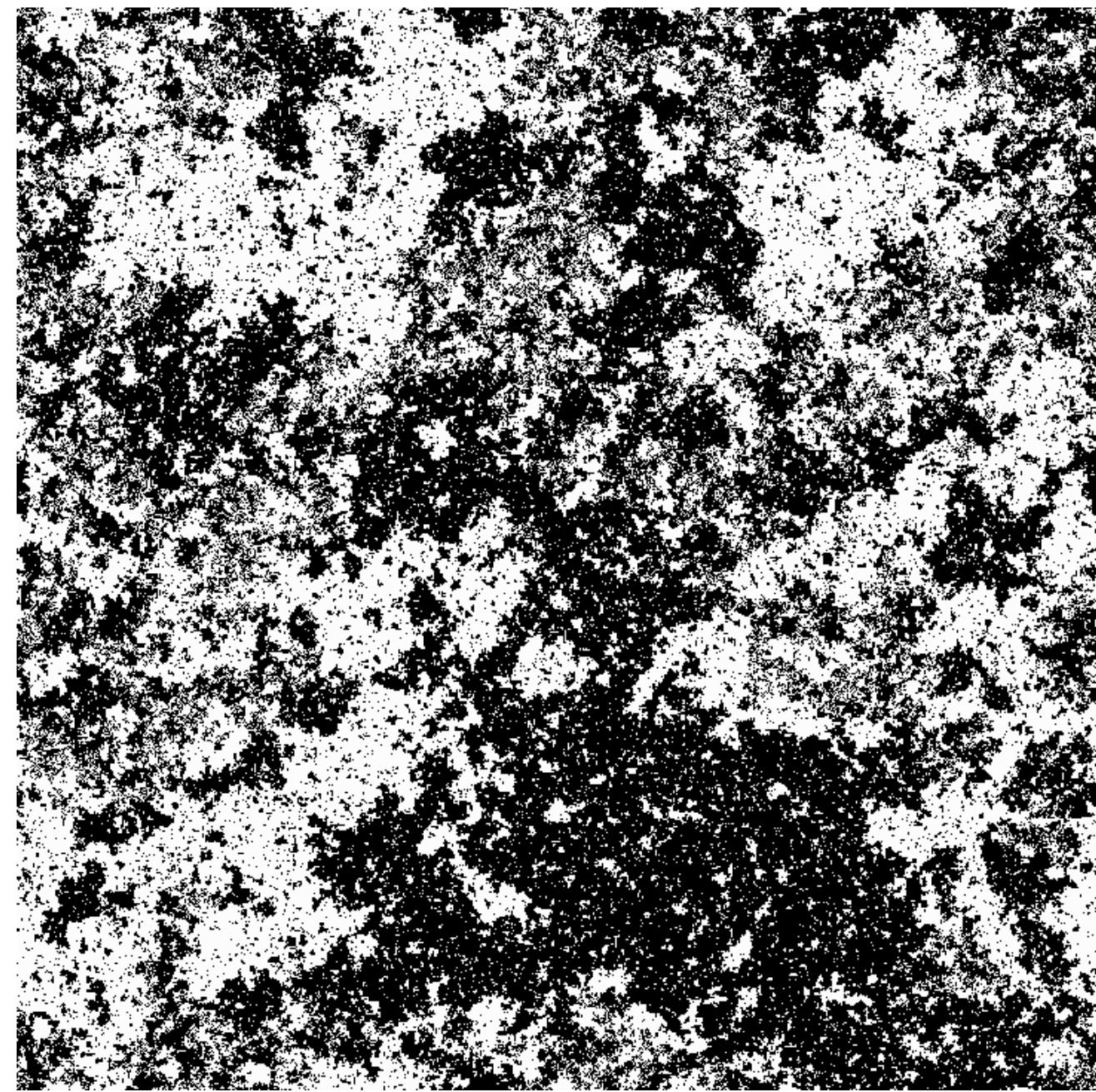
Explanation for the “Fractal-like” structure?

Properties of Critical phenomena: Scale Invariance



Bond percolation

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Ising model

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Explanation for the “Fractal-like” structure?

Close to criticality:

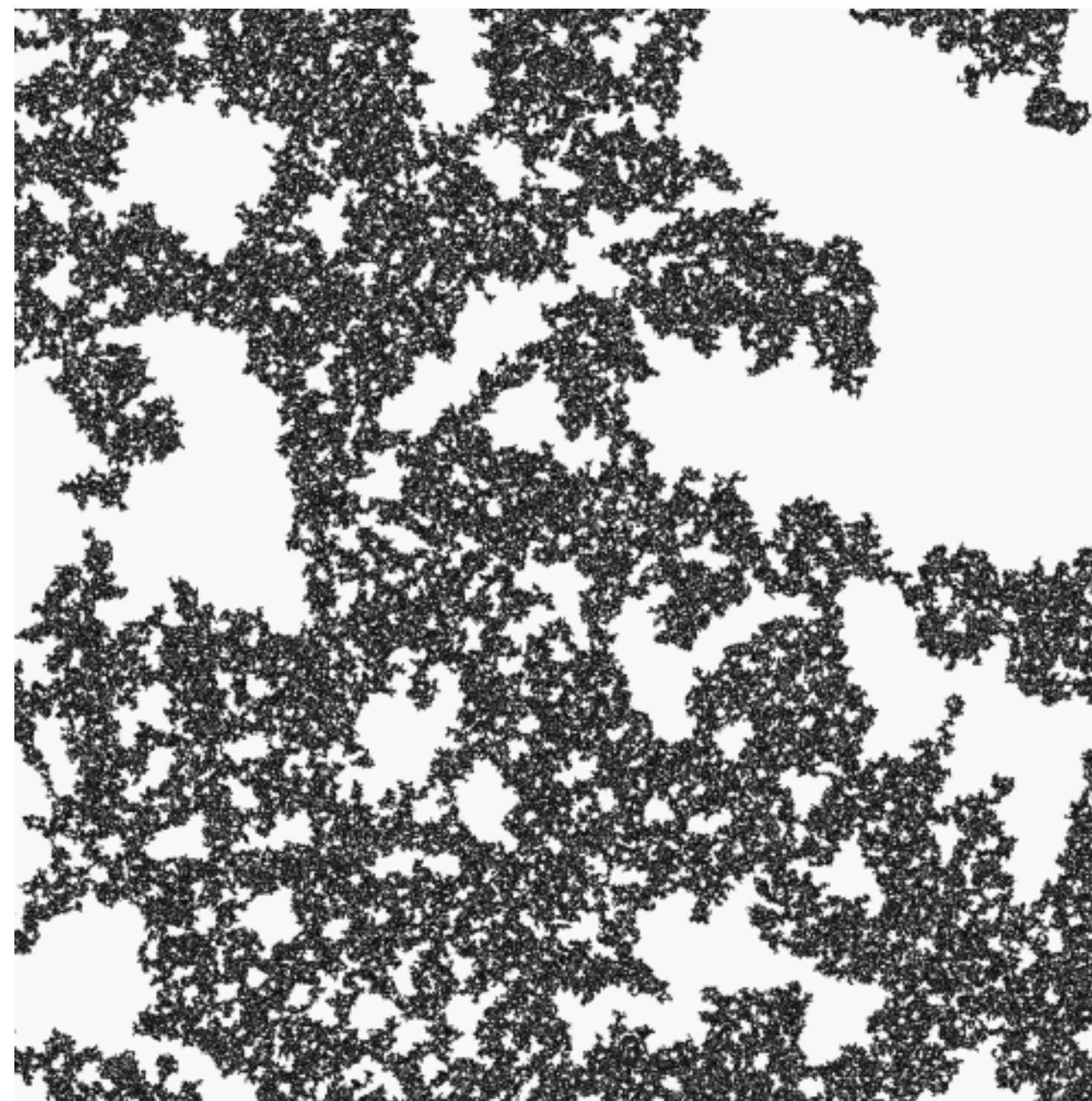
Diverging correlation length —> percolating cluster

Cluster of all sizes

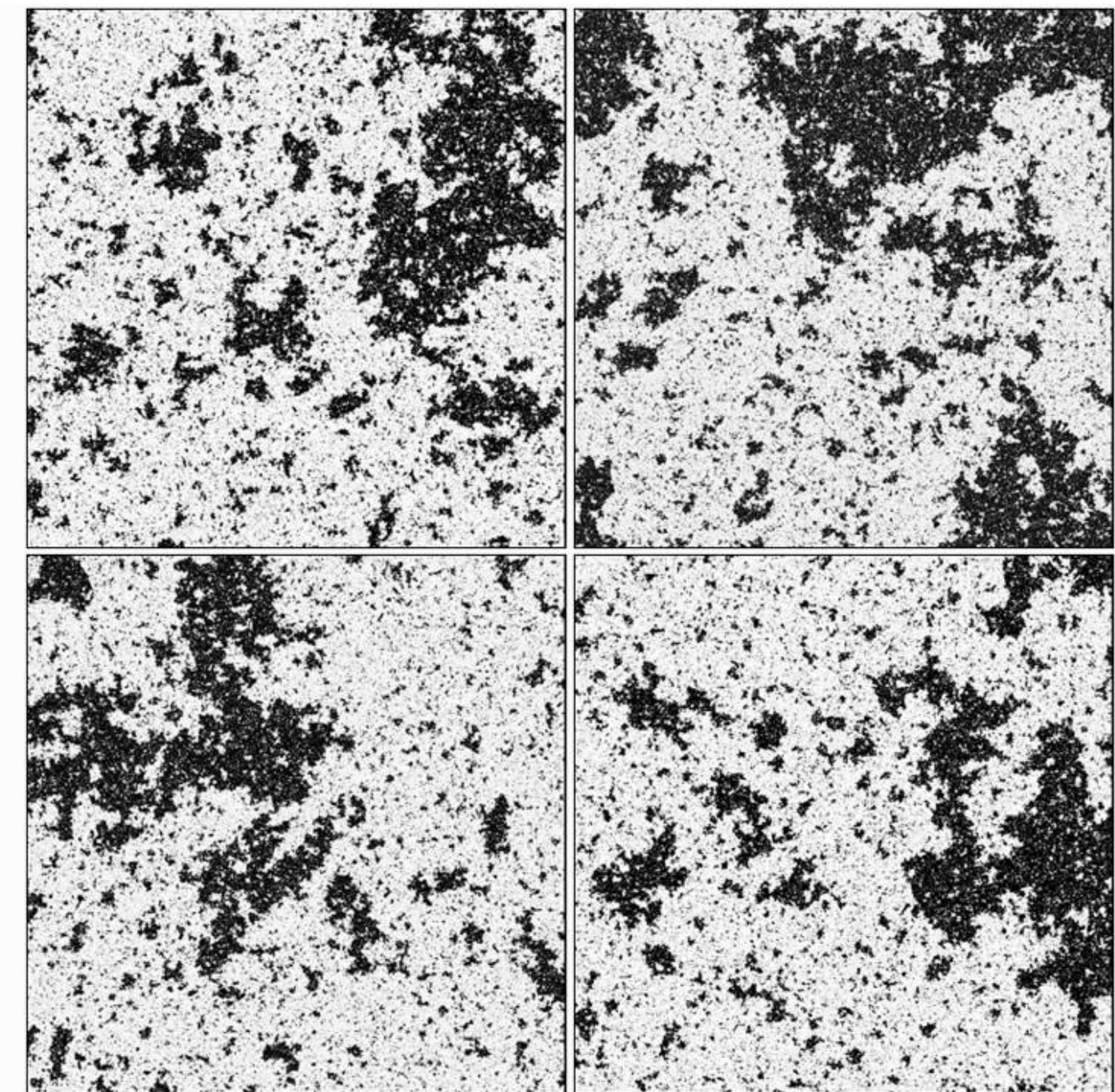
Diverging correlation length

Fluctuation of the magnetization at all scales

Scale invariance: the system looks the same at all scales! (i.e. when zooming in)



Scale invariance of critical percolating cluster



Scale invariance in the critical Ising Model

Self-similarity and Fractals

Fractals in nature?

Fractal dimension

Fractals

Self-similar object

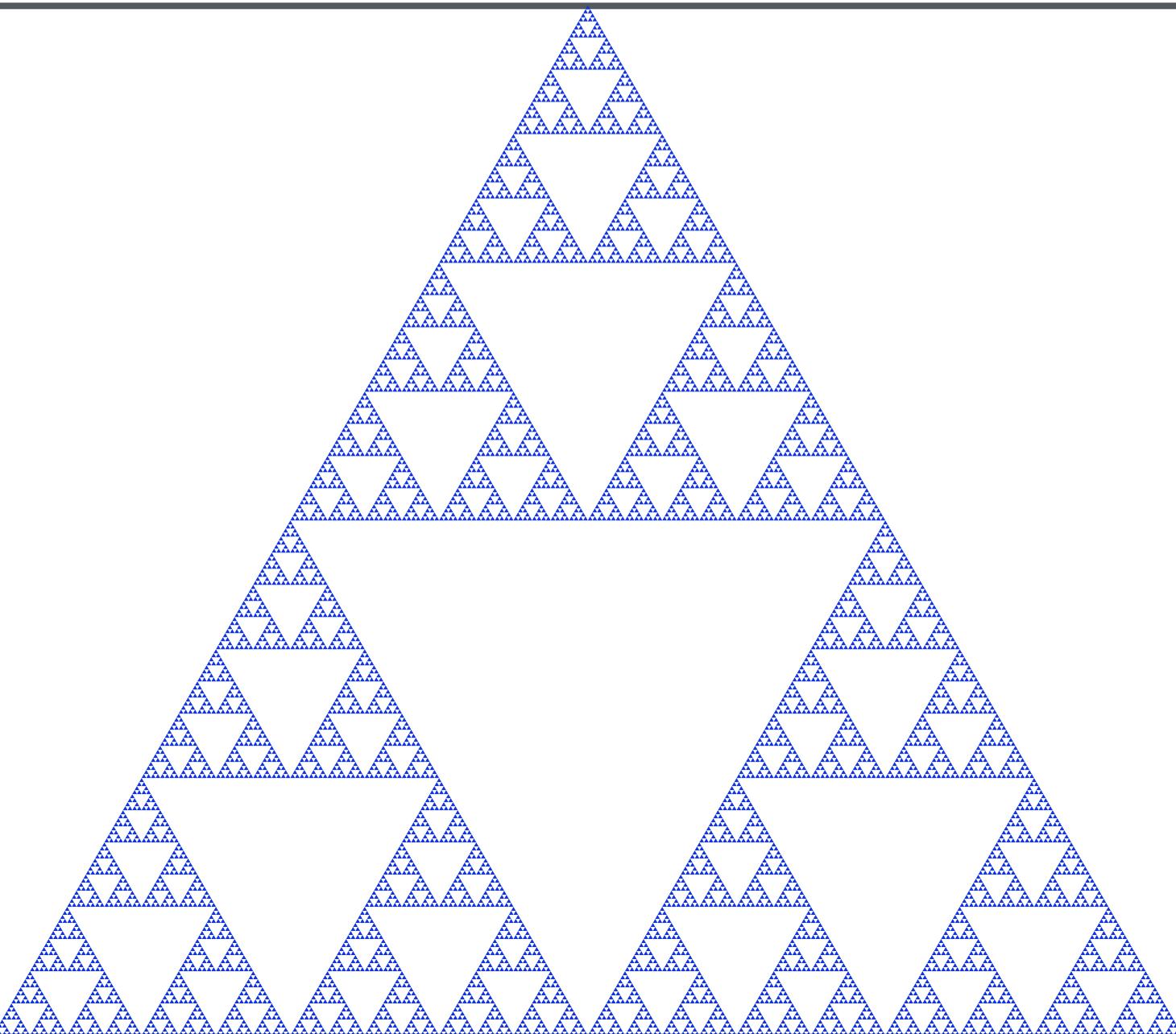
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It is a typical properties of **fractals**

Example: the Sierpiński triangle

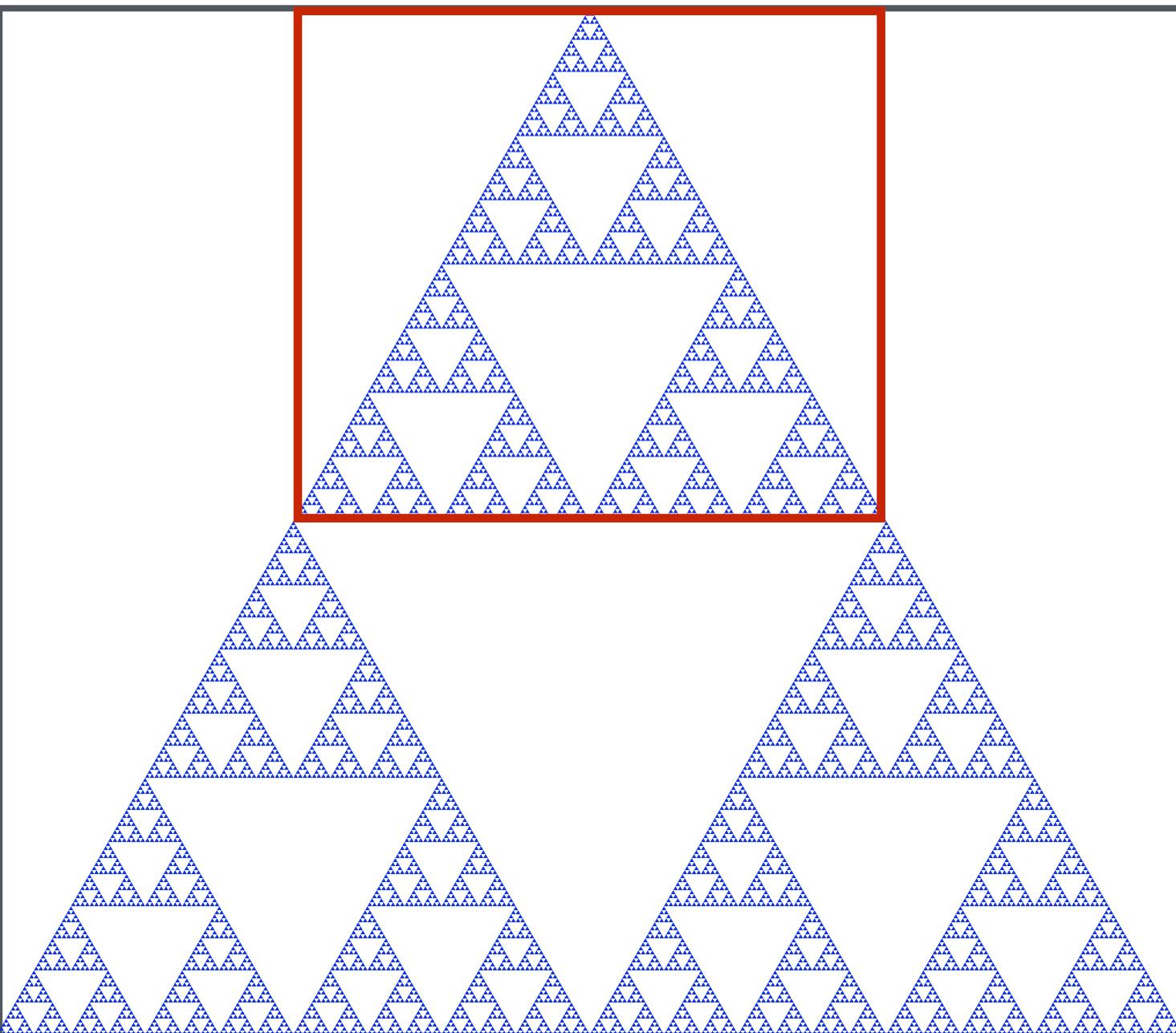


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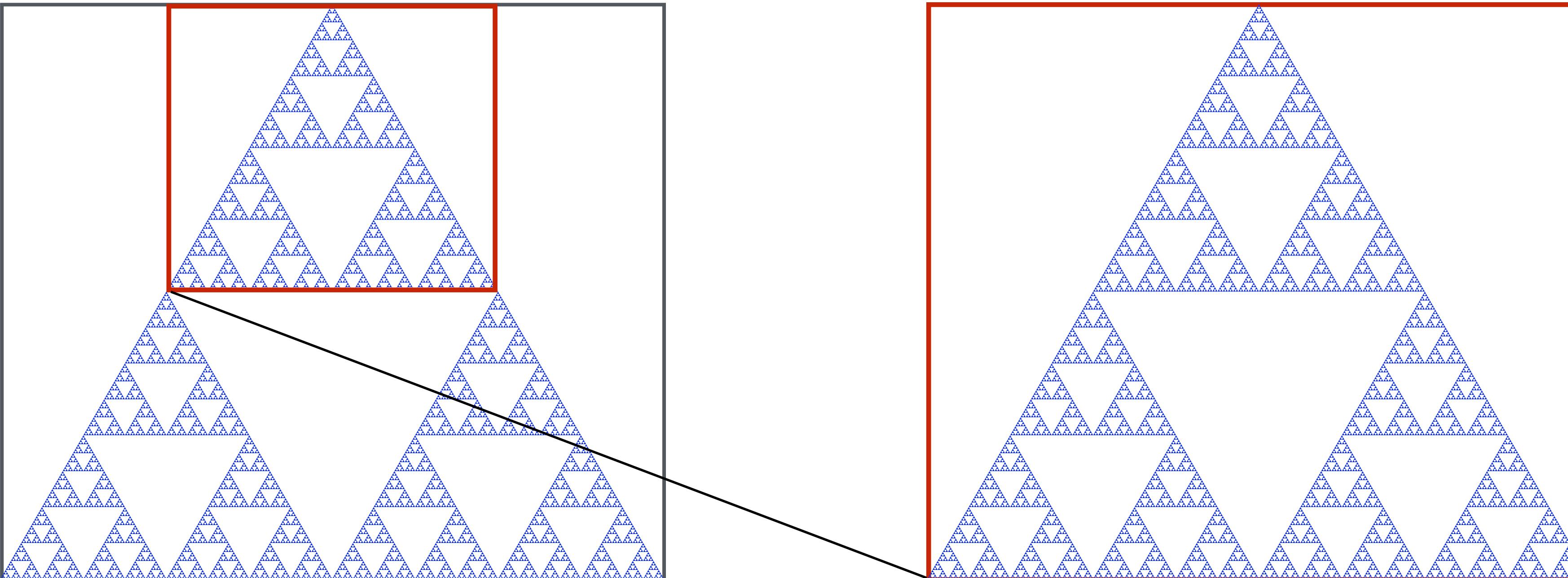


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Ex. of Fractals/Self-similarity in nature?

Many real world elements are **statistically self-similar**:



Frost crystals

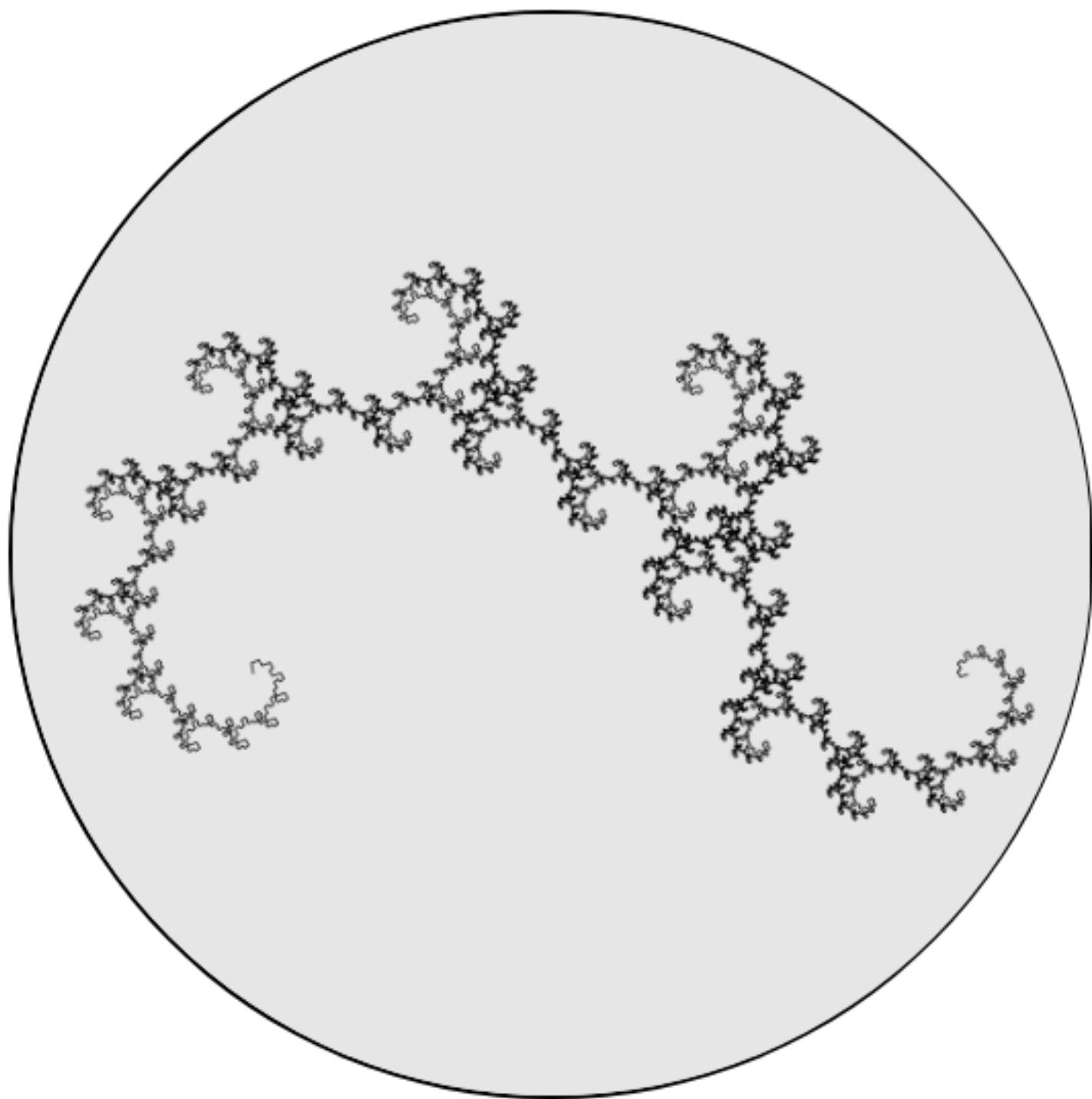


Romanesco broccoli



High-voltage breakdown within a
block of acrylic glass

Examples of Models

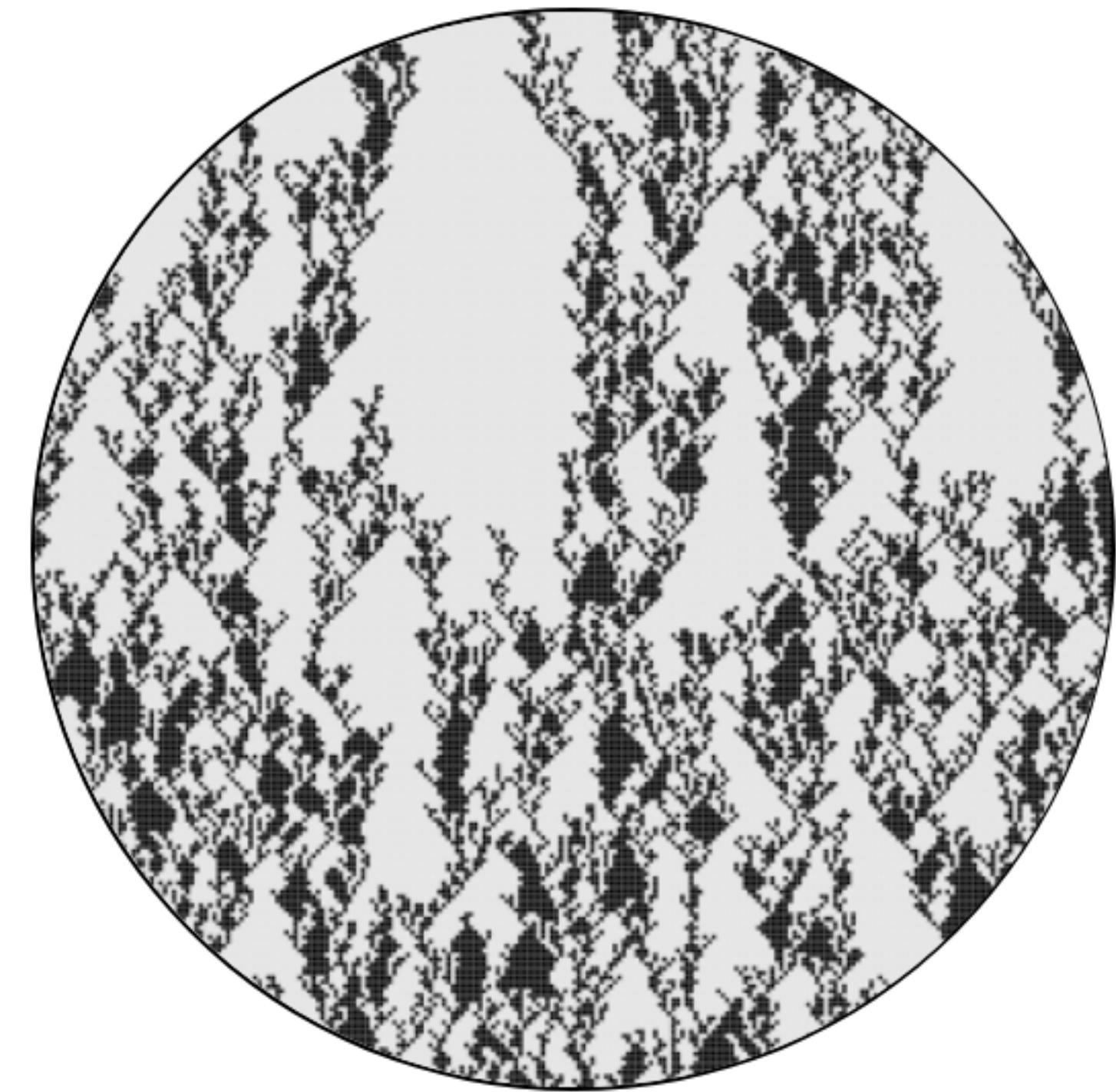


Iteration process

in which elements of a structure are replaced by a smaller version of the whole structure



Lindenmayer Systems
Statistical iterative system:
how fractal patterns observed in natural systems, particularly structural properties of some plants

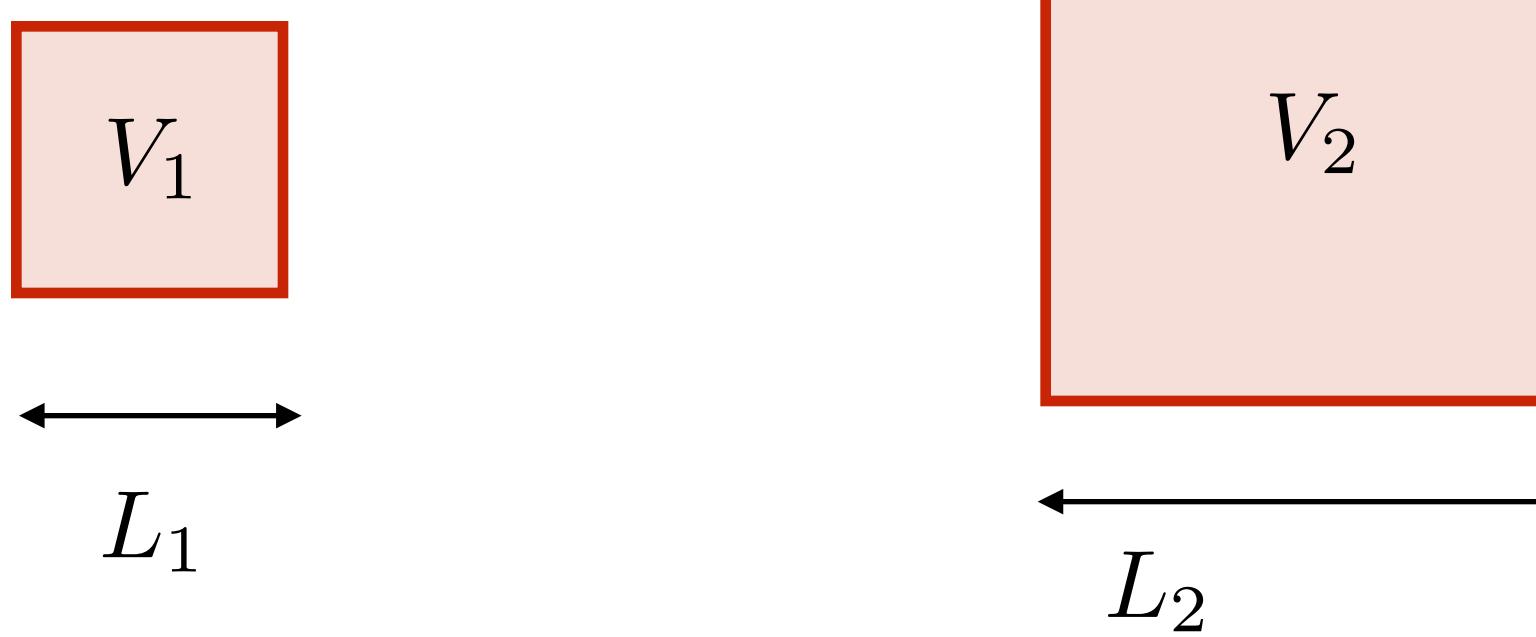


Stochastic cellular automaton
generating fractal growth patterns

Fractal dimension

Usual objects: expect volume (or mass) scales as

$$V \propto L^d \quad \text{where } d = \text{dimension of the object} \\ = \text{is an integer}$$

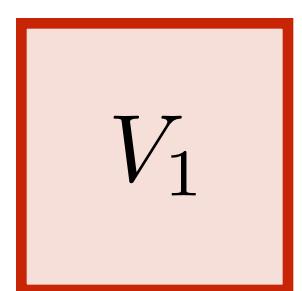


$$\rho = \frac{V_1}{L_1^d} = \text{constant}$$

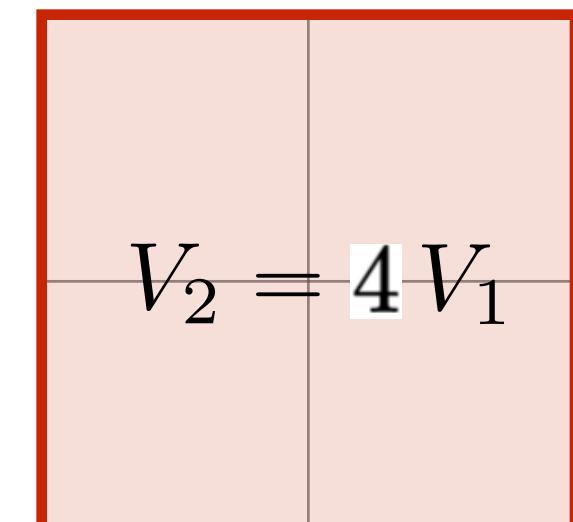
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$$L_1$$



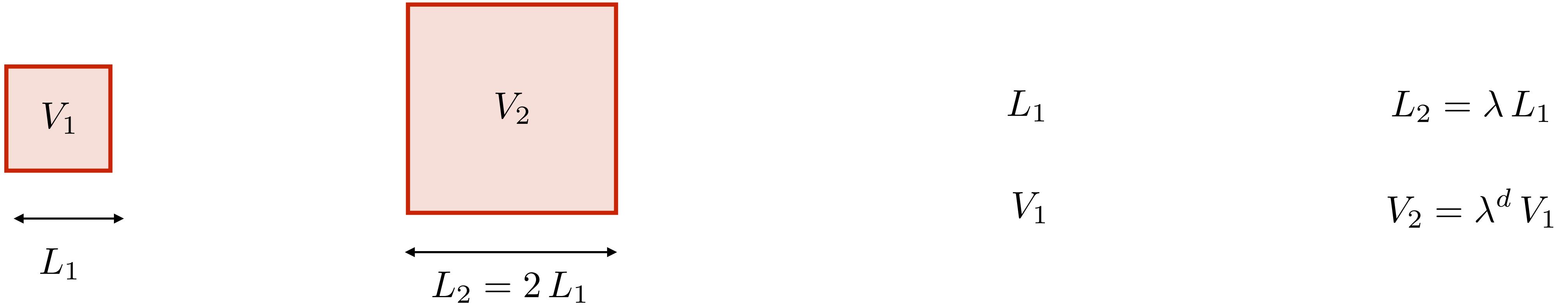
$$L_2 = 2L_1$$

Density is independent of the length scale at which it is measured

$$\rho = \frac{V_1}{L_1^d} = \frac{V_2}{L_2^d} = \text{constant}$$

Fractal dimension

Usual objects: usually the density is independent of the length scale at which it is measured: $\rho = \frac{V_1}{L_1^d} = \frac{V_2}{L_2^d}$



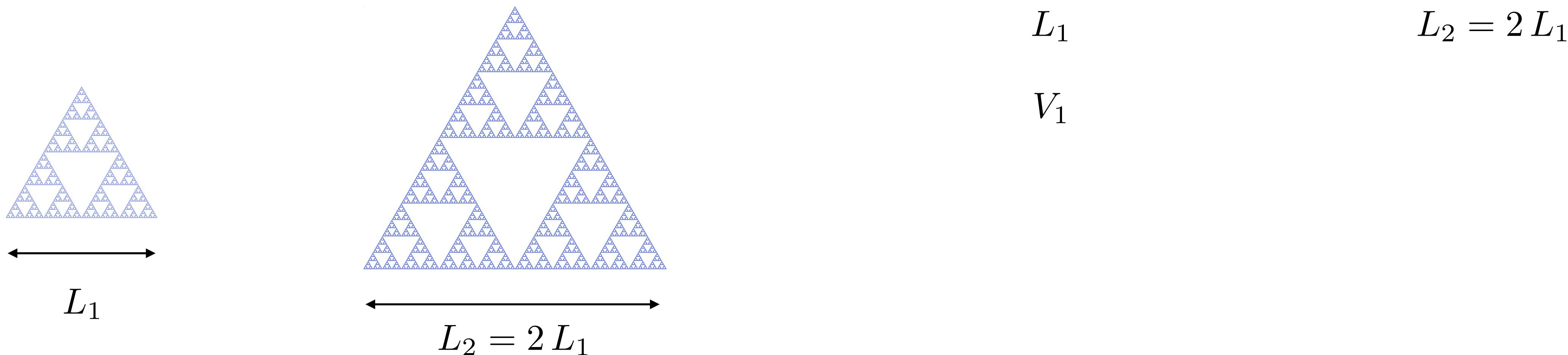
Fractal objects: this idea [$V_2 = \lambda^d V_1$] is extended to fractal objects —> defines the notion of **fractal dimension**

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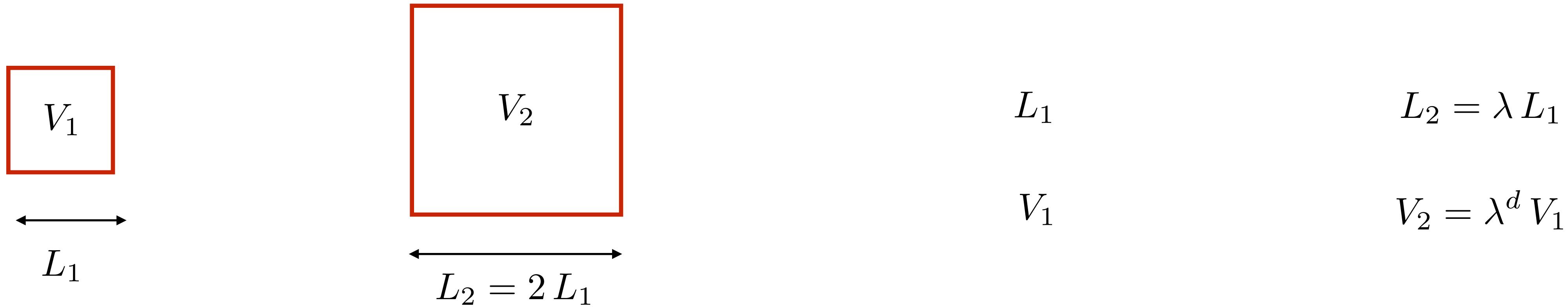


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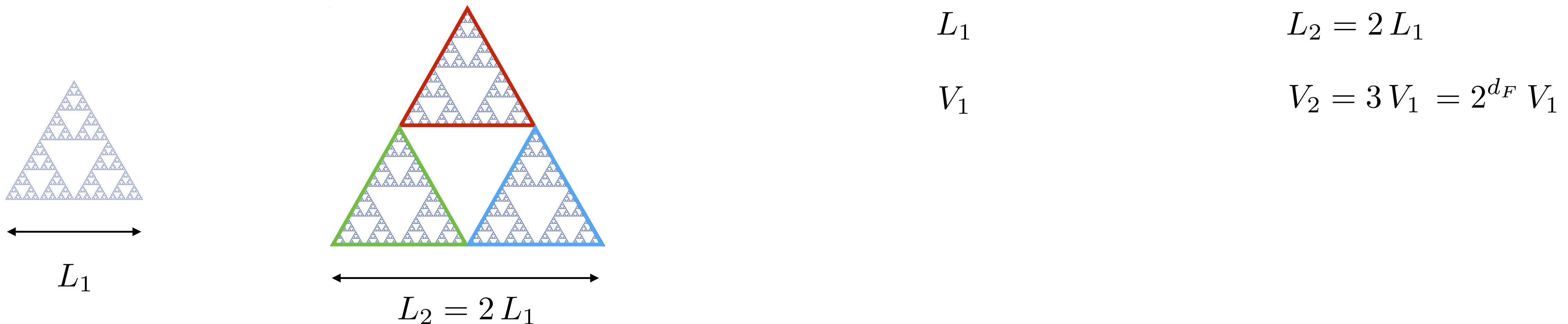


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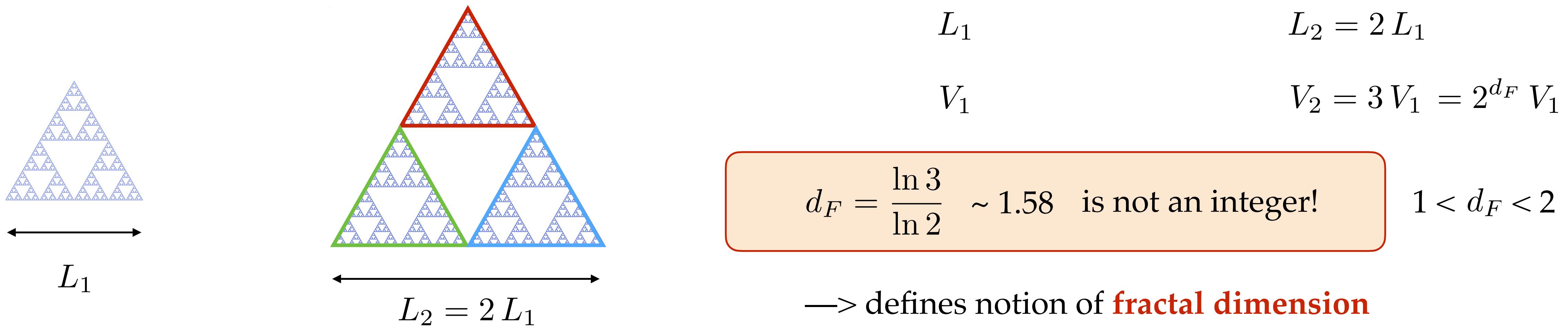


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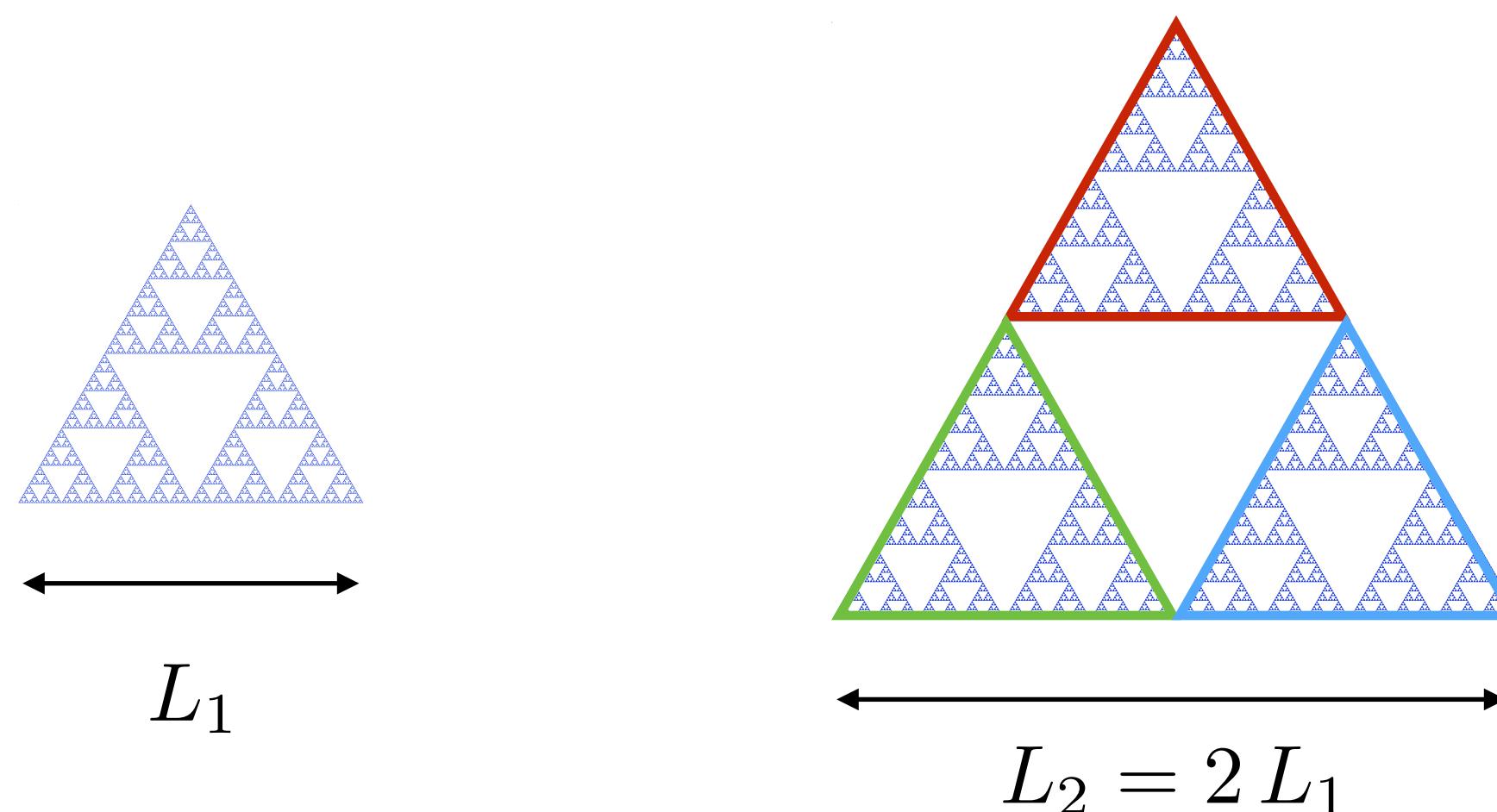


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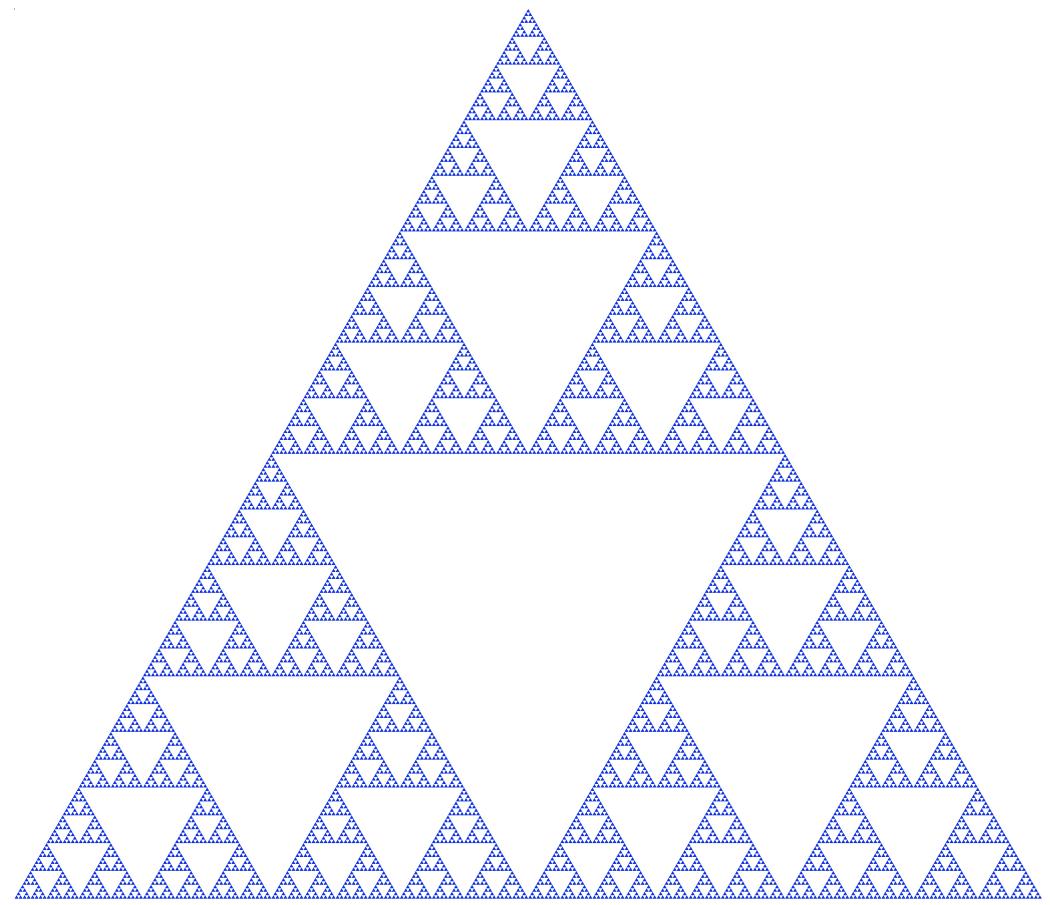


$$\rho_2 = \frac{V_2}{L_2^d} = \frac{\lambda^{d_F}}{\lambda^d} \frac{V_1}{L_1^d} = \frac{1}{\lambda^{d-d_F}} \rho_1 \quad 1 < d_F < 2 \Rightarrow d$$

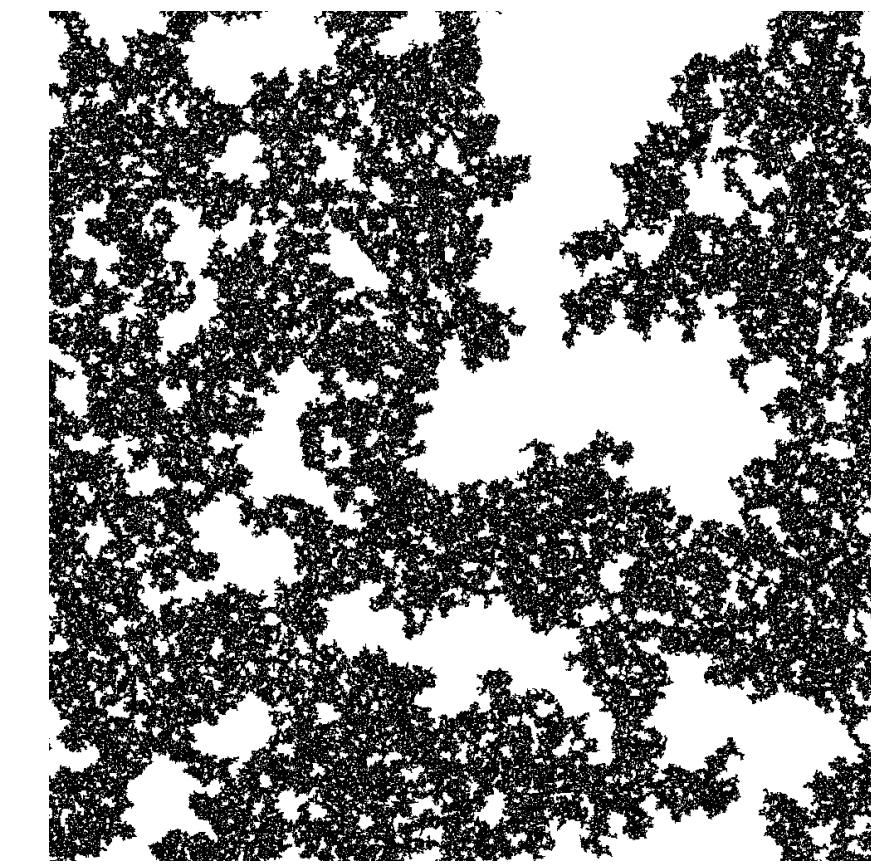
Density depends on the scale! $\rho_2 = \frac{1}{\lambda^{d-d_F}} \rho_1$
Decrease for larger lambda

At criticality: Statistical Self-similarity and Scale Invariance

Exact self-similarity

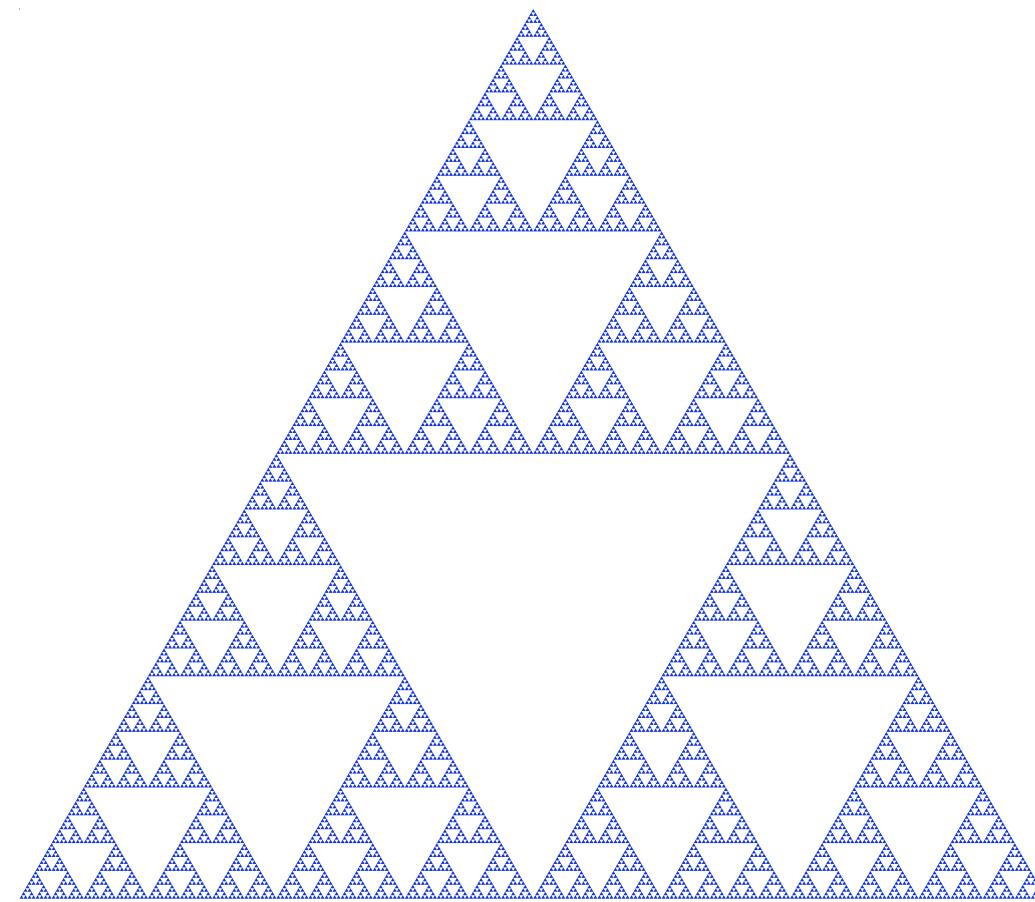


Scale invariance



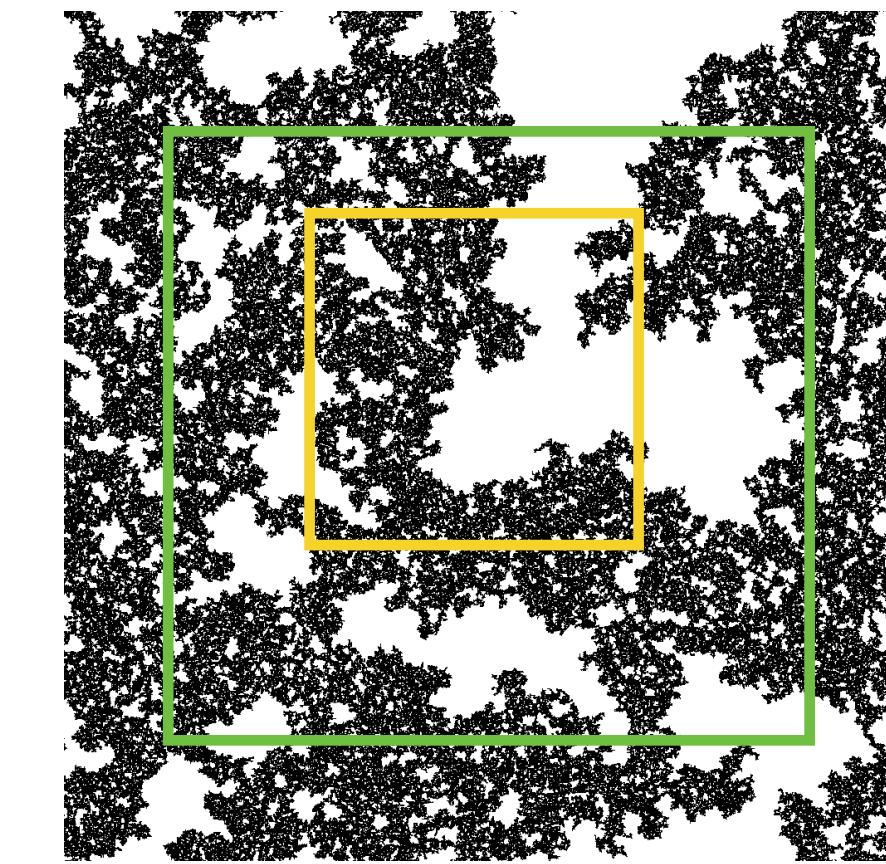
At criticality: Statistical Self-similarity and Scale Invariance

Exact self-similarity



Self-similar only for specific values of scaling factor λ

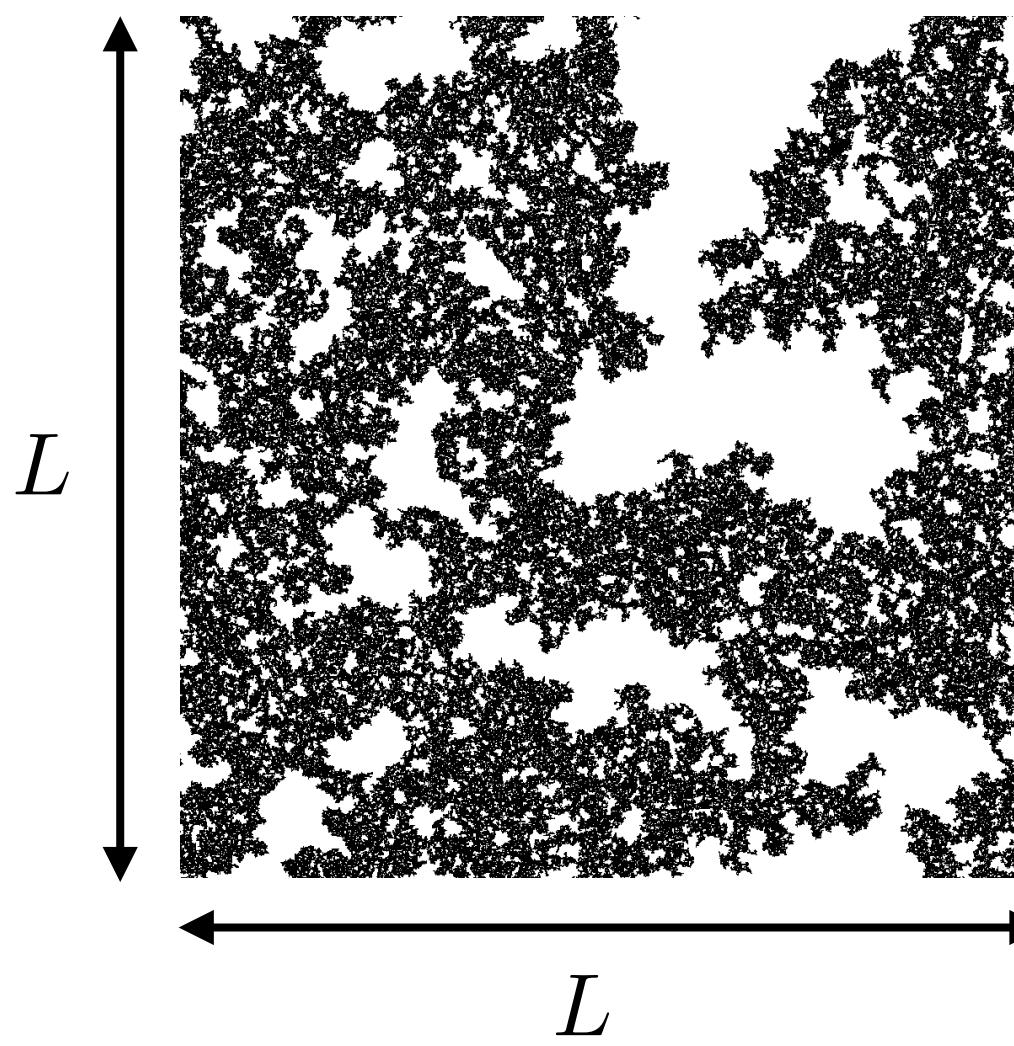
Scale invariance



There is a small piece of object that is similar to the whole,
at any choice of the scaling factor λ

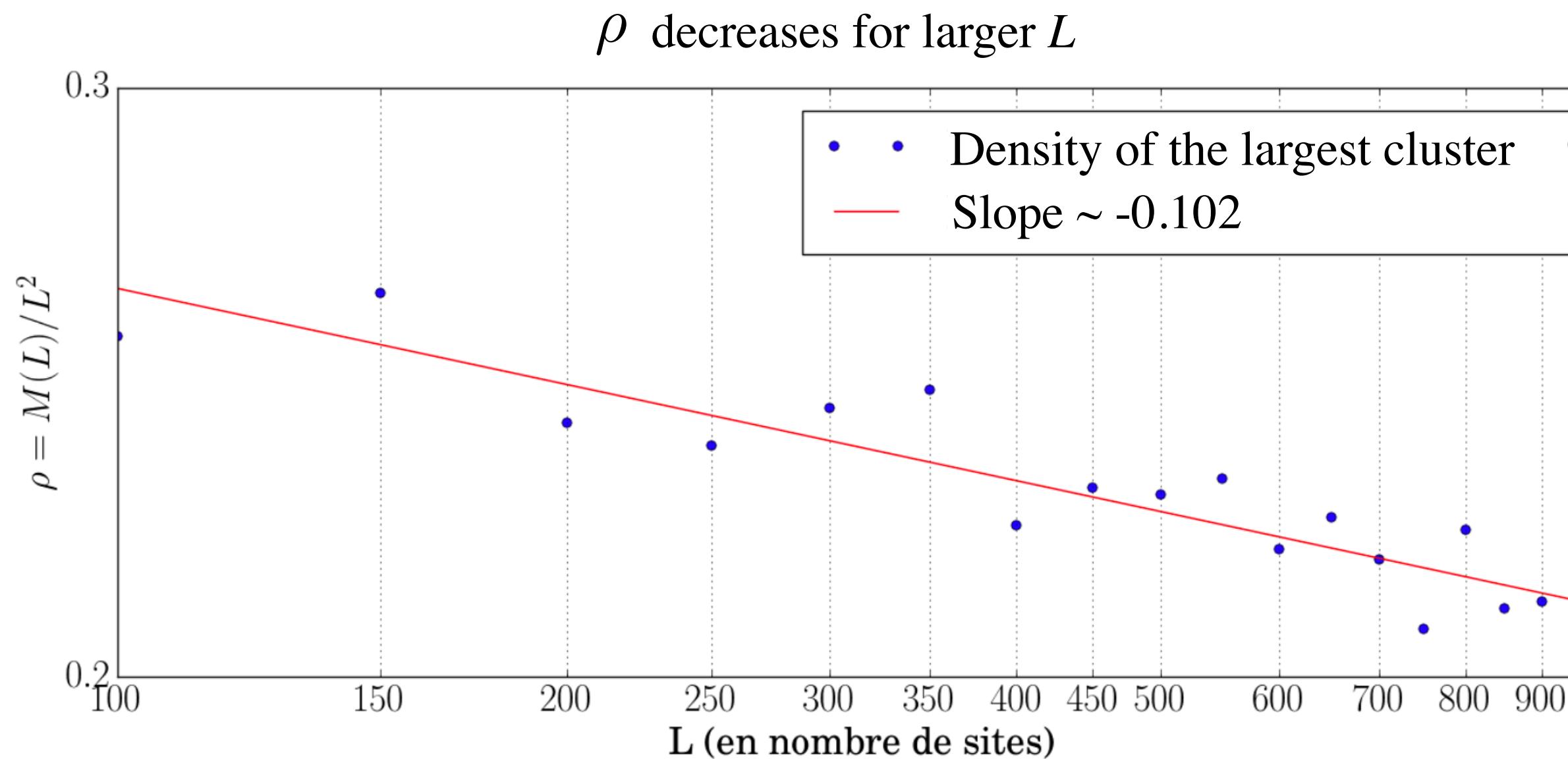
is self-similar for all choices of scaling factor λ

At criticality: Scale Invariance



Site percolation
on a square lattice

at $p = p_c \sim 0.59$



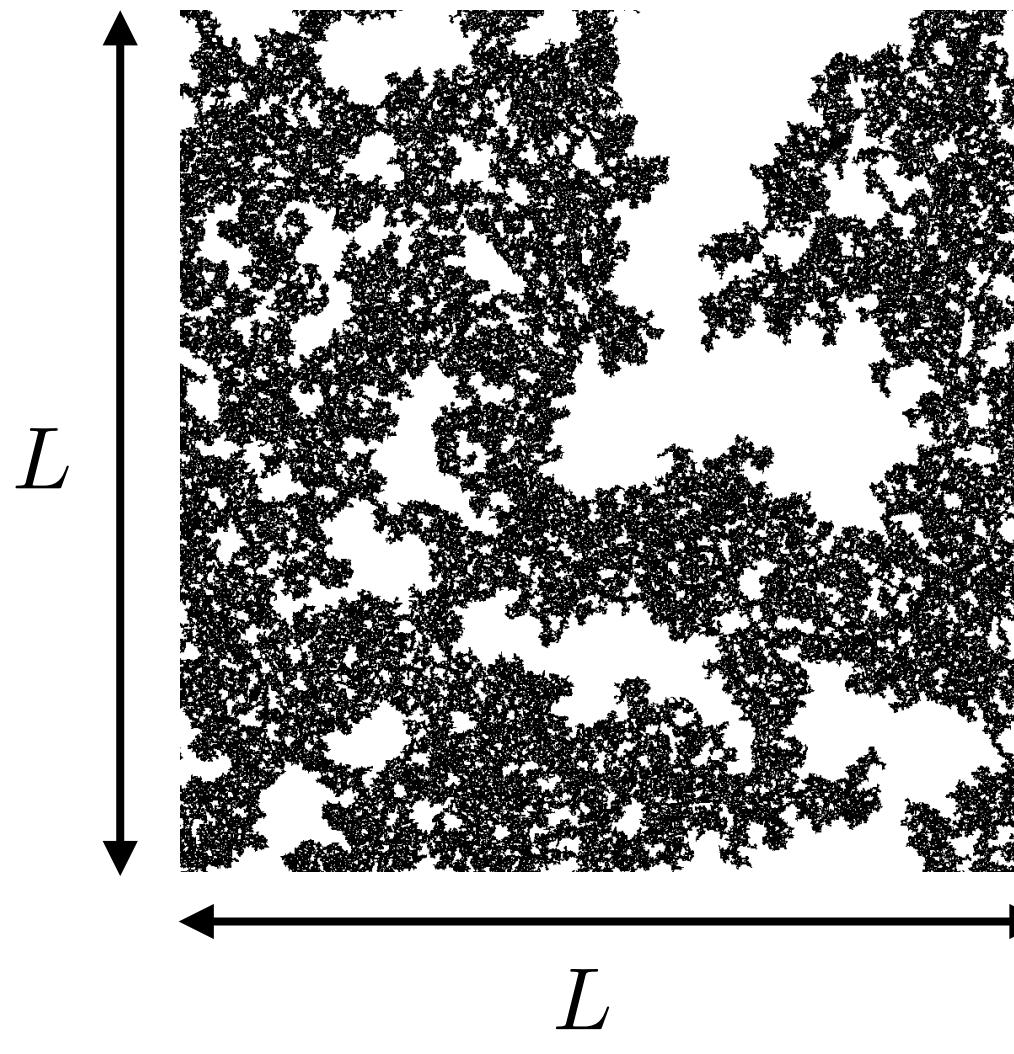
ρ decreases for larger L

$$\frac{\rho_2}{\rho_1} = \left(\frac{L_2}{L_1} \right)^{-(d-d_F)}$$

$$\log \rho = -(d - d_F) \log(L) + \text{const.}$$

$$d_F \simeq -0.102 + d \simeq 1.898$$

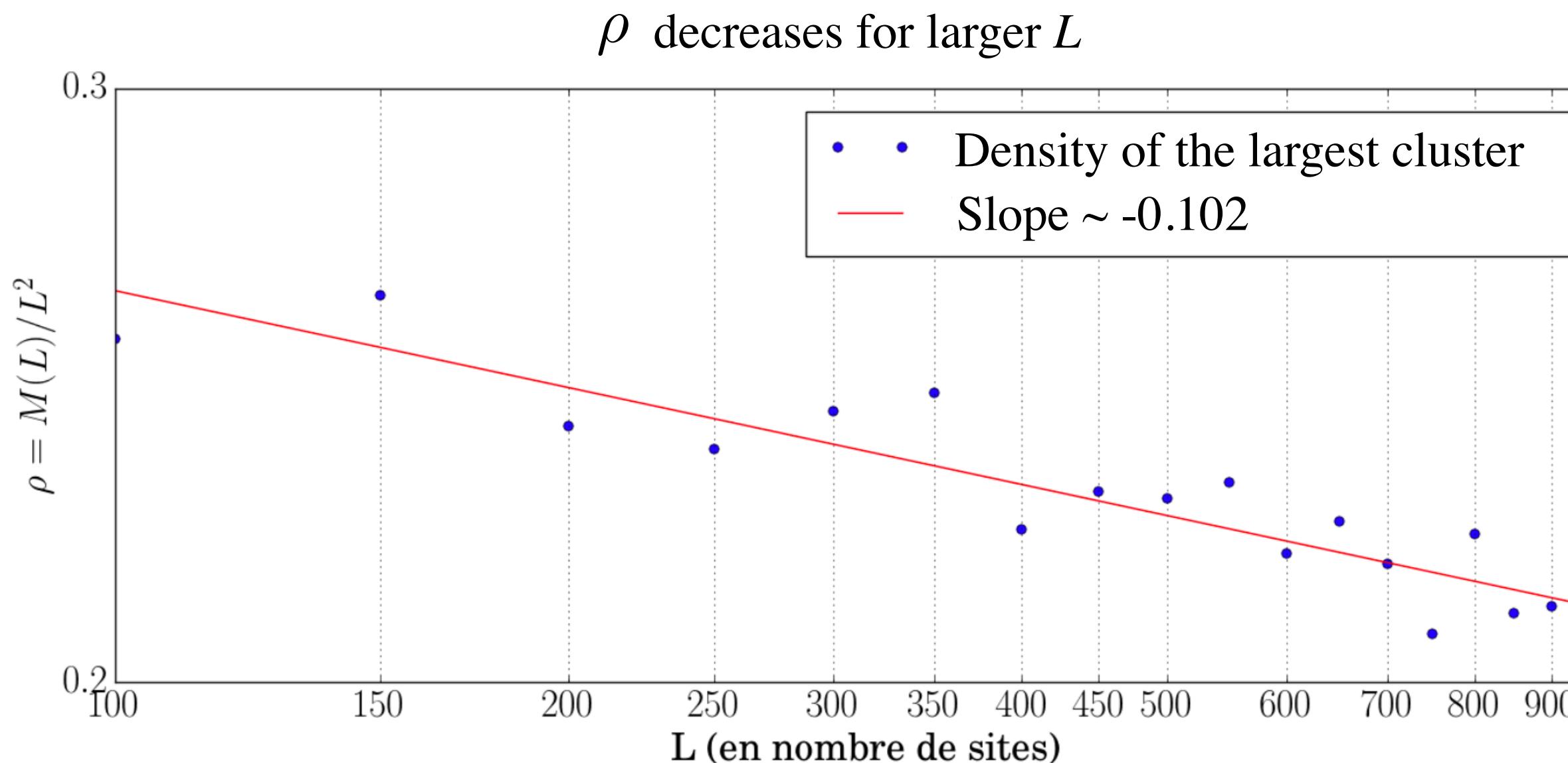
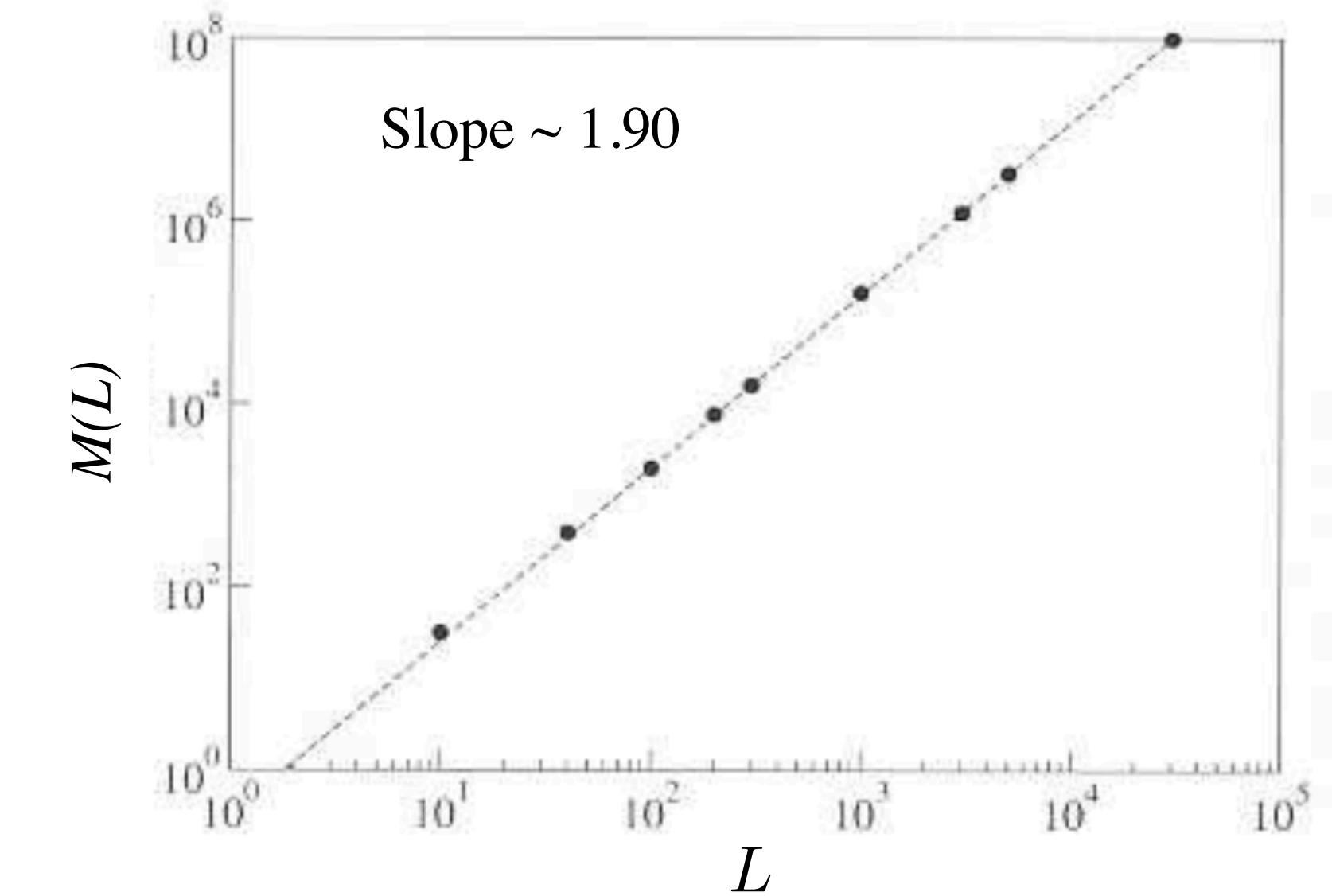
At criticality: Scale Invariance



Site percolation
on a square lattice
at $p = p_c \sim 0.59$

$$M(L) \propto L^{d_F}$$

$d_F \simeq 1.90$



ρ decreases for larger L

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Random walks

Scale-invariance

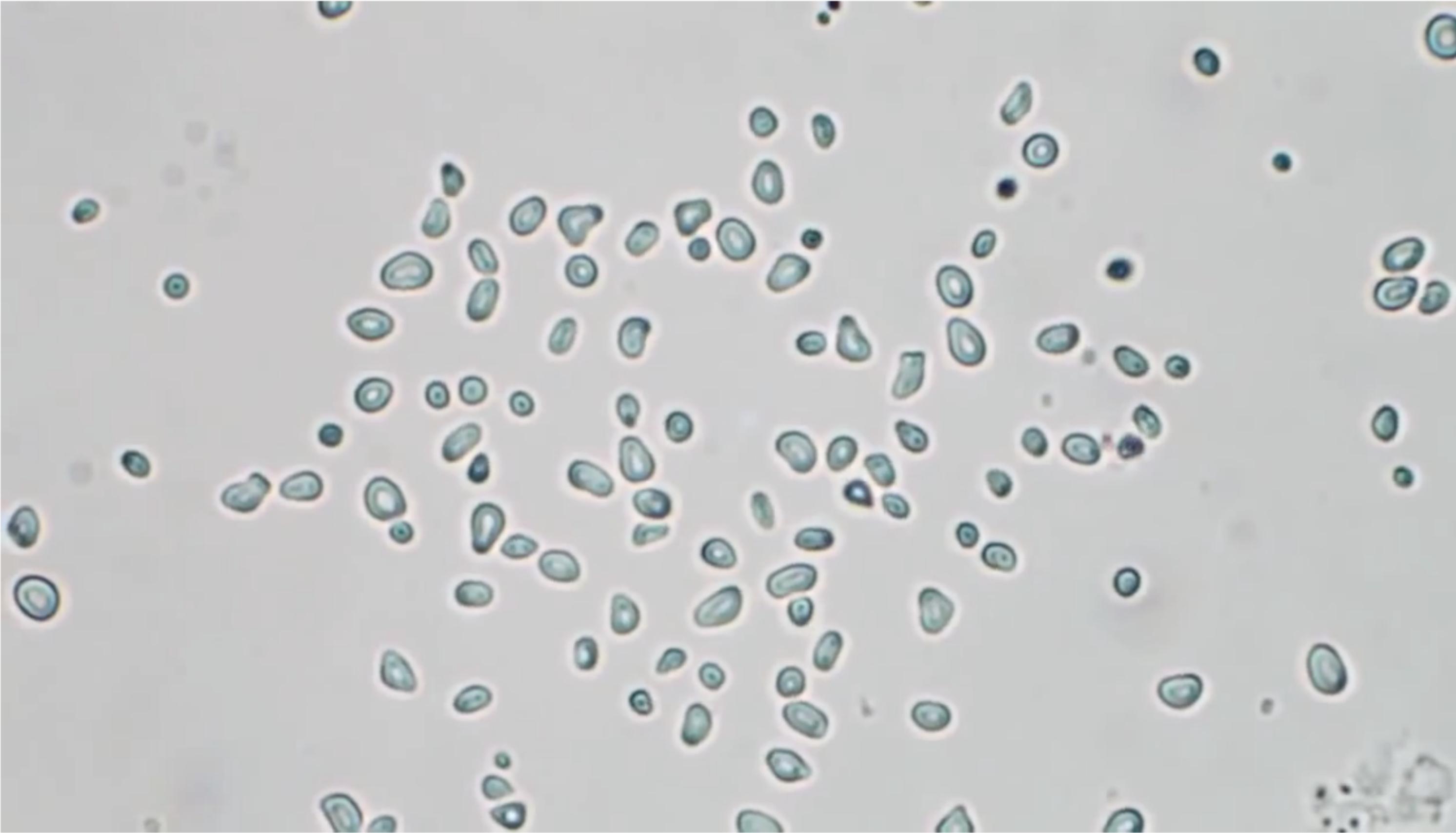
Universality

Fractal dimension

**Random walk,
Self-Similarity, Universality,
Fractal dimension**

Brownian motion

1827: Robert Brown observed irregular motion of small pollen grains suspended in water

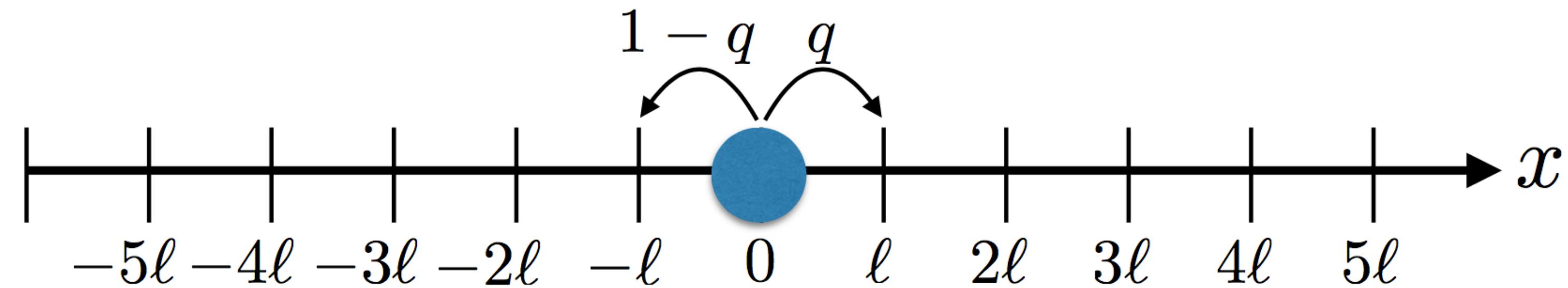


1905-06 Albert Einstein, Marian Smoluchowski: microscopic description of Brownian motion and relation to diffusion equation.

Random walks

Random walks: paths that take successive steps in random directions.

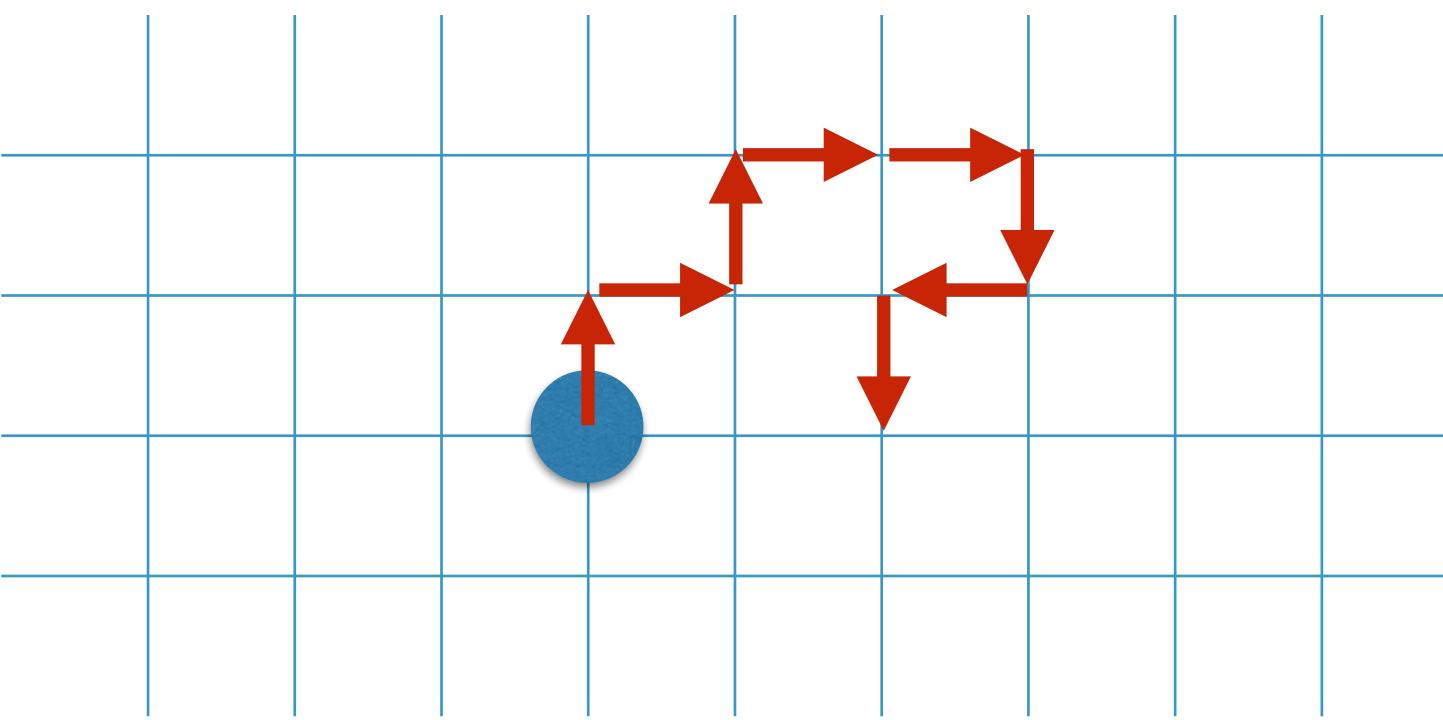
Ex. 1D



Ex. 2D

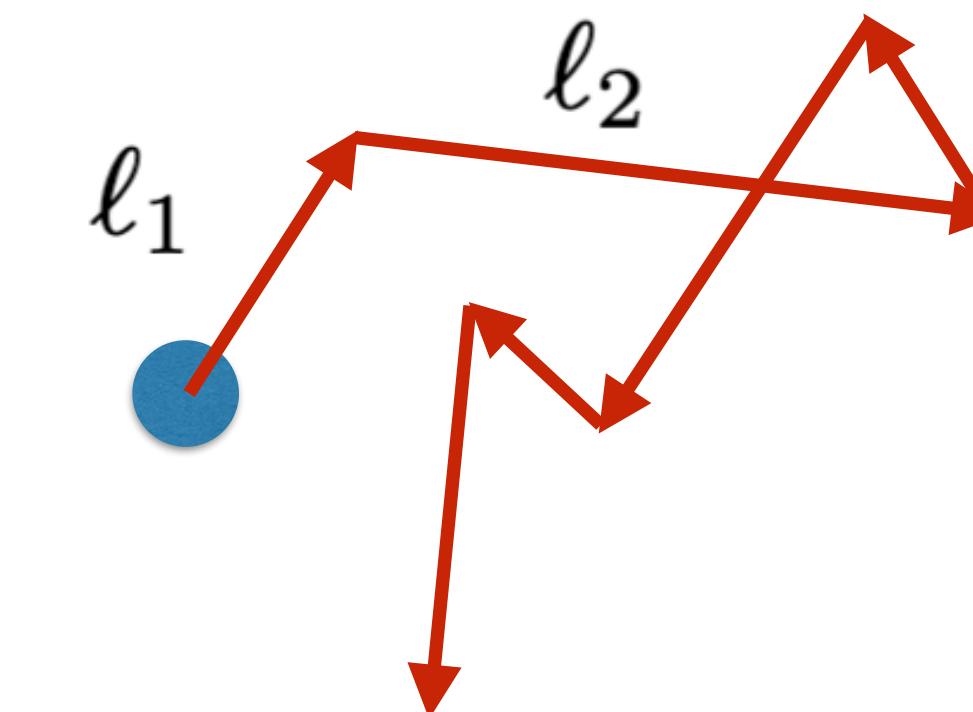
Ex. Random walk on a 2D square lattice

1/4 chance to go in any of the 4 directions



Ex. Random walk in the continuous 2D plane

Choose a random direction uniformly
Jump of length ℓ sampled from a continuous distribution $P(\ell)$

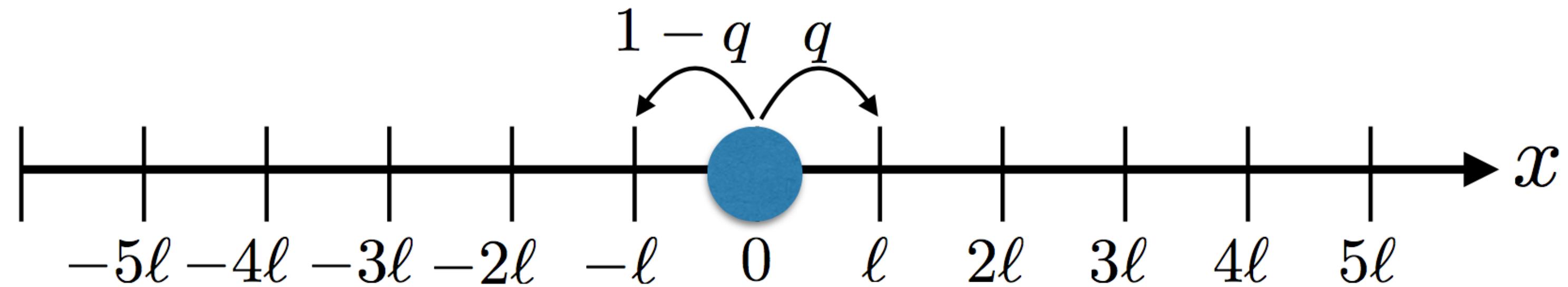


Random walks

Random walks: paths that take successive steps in random directions.

What can we say about this system?

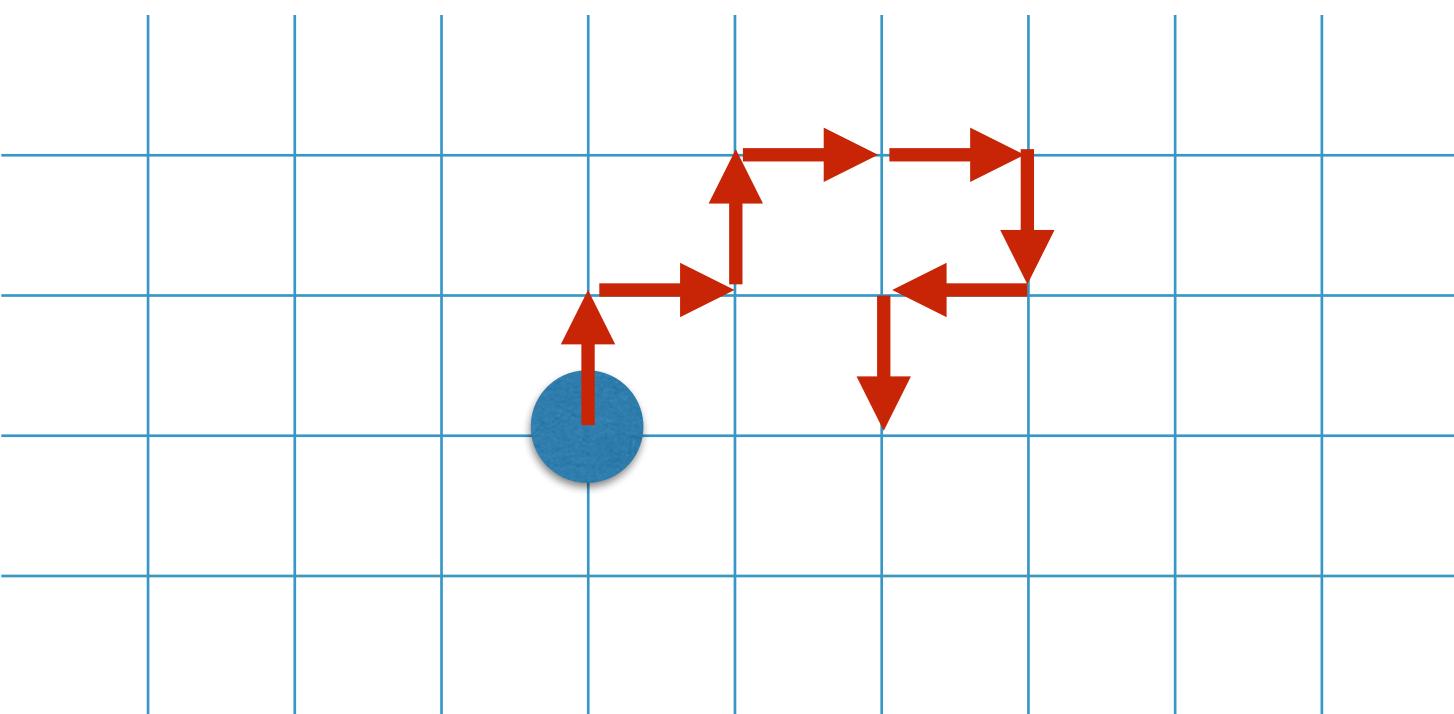
Ex. 1D



Ex. 2D

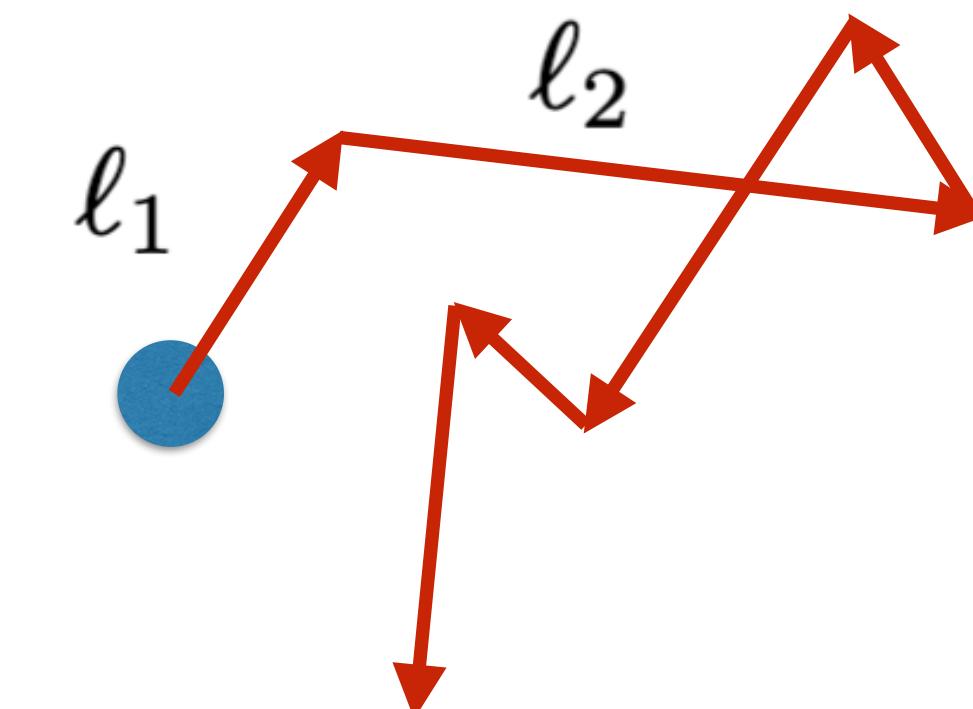
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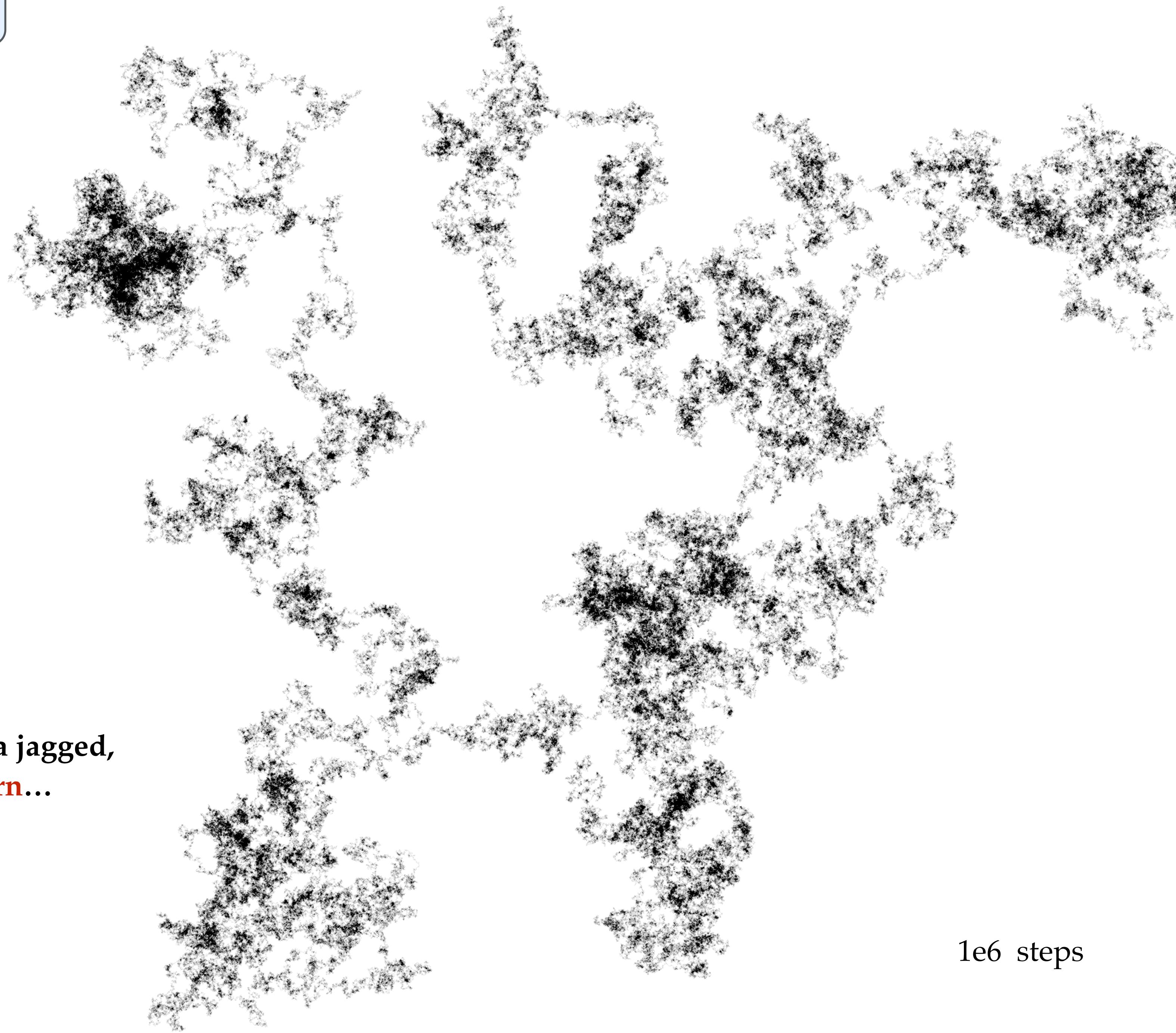


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Random walks

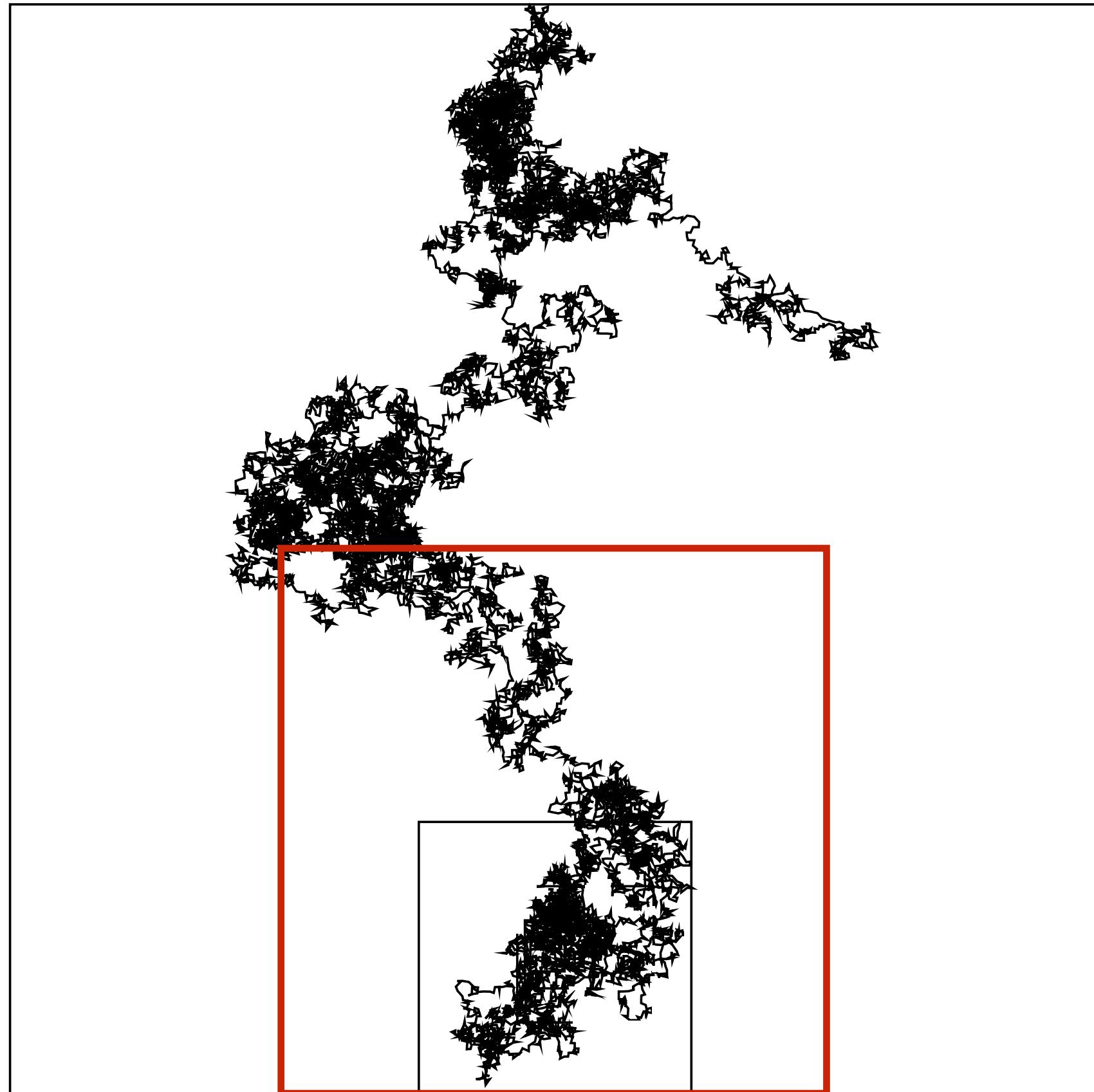


Random walks form a jagged,
fractal-like pattern...

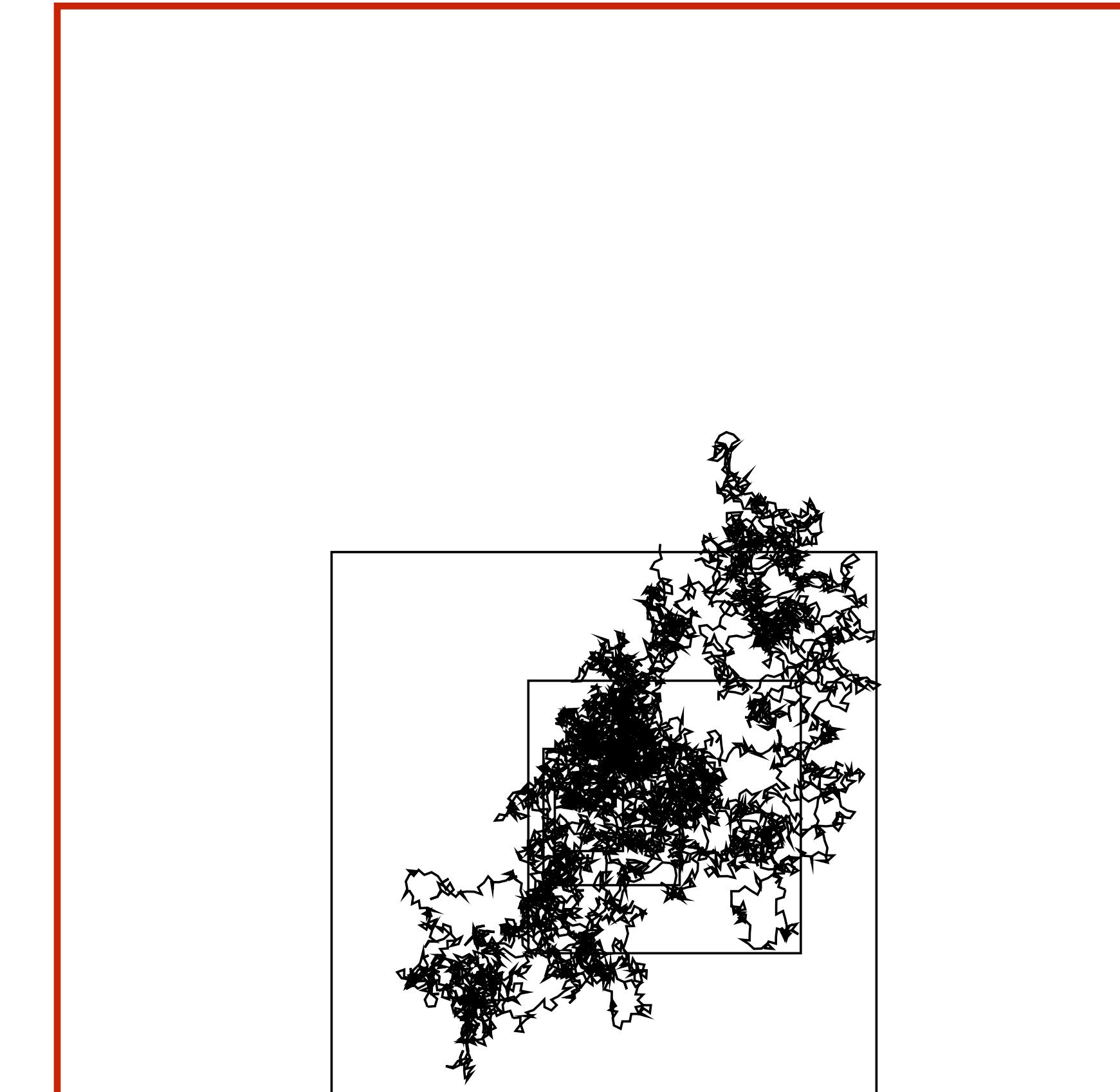
1e6 steps

Scale-invariance

Scale invariance of random walks



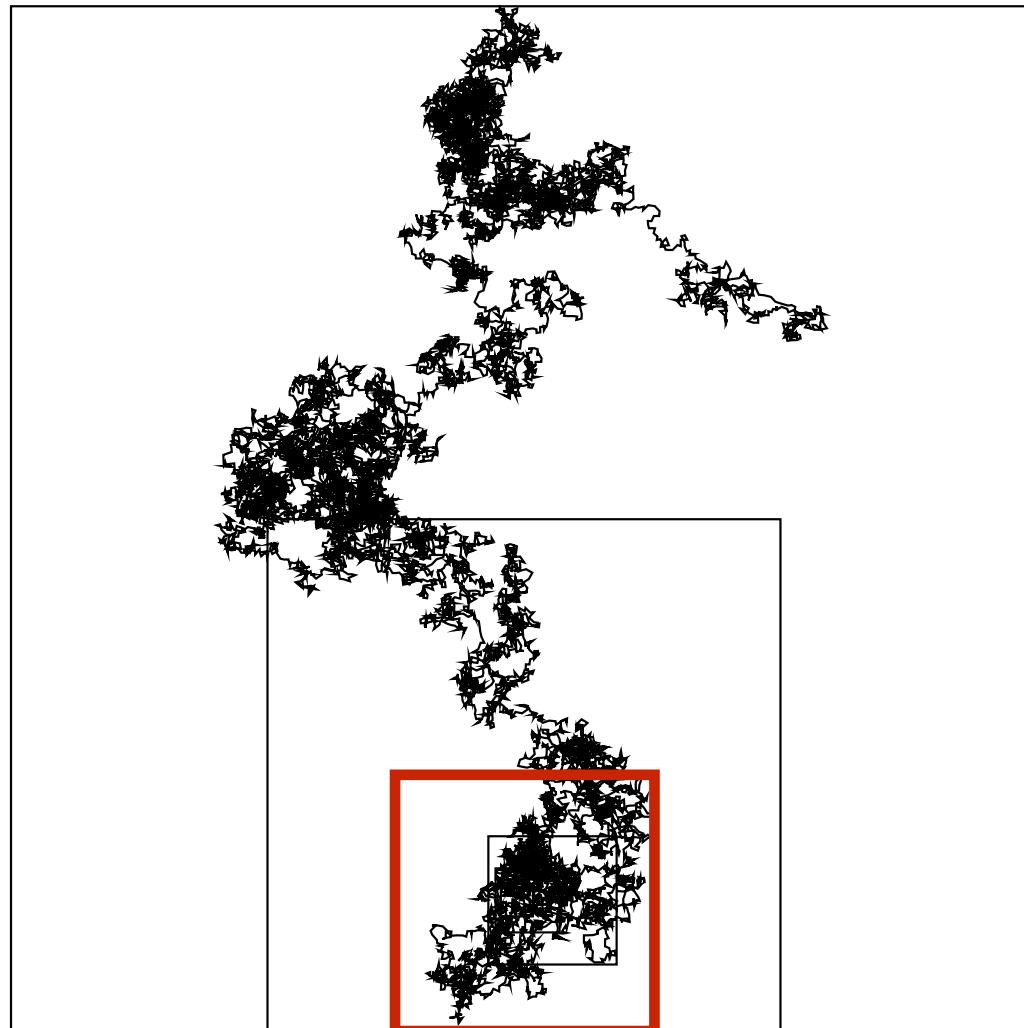
32 000 steps



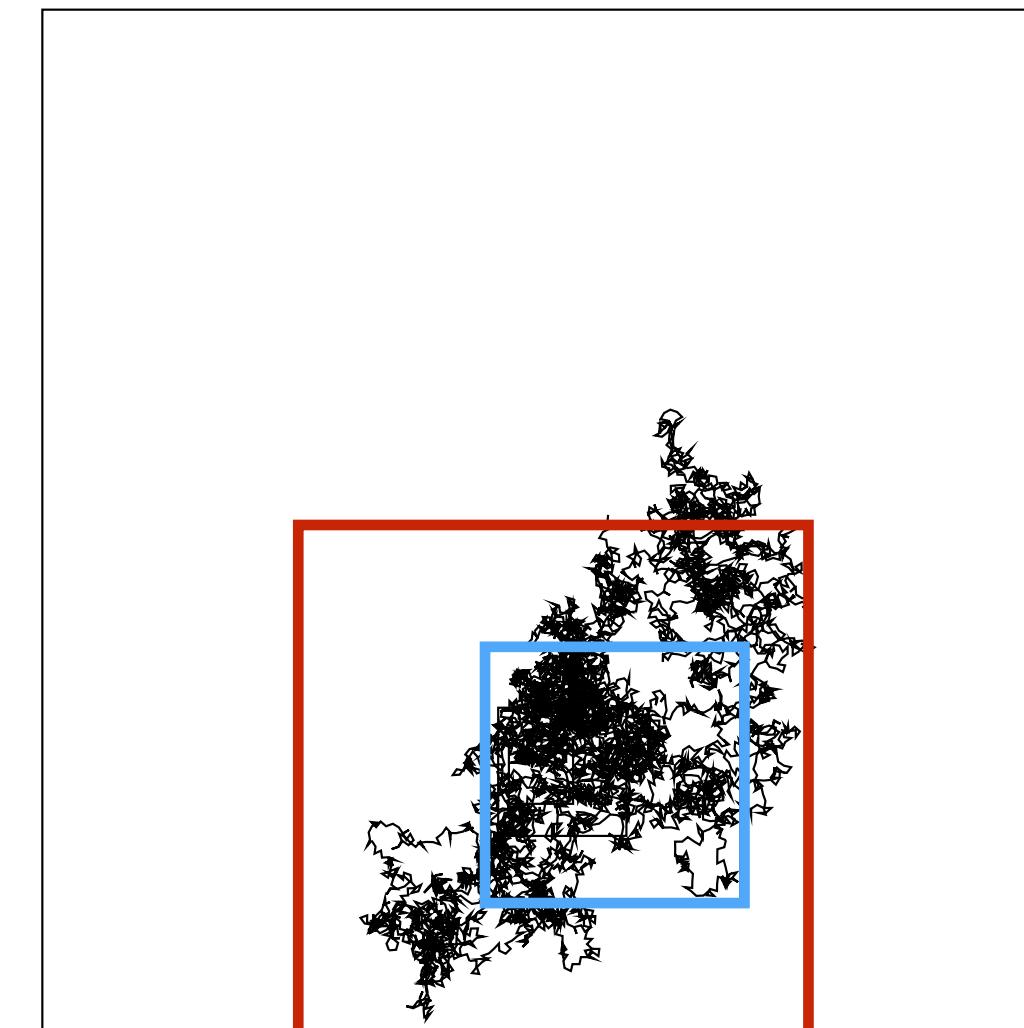
First 8 000 steps

Scale-invariance

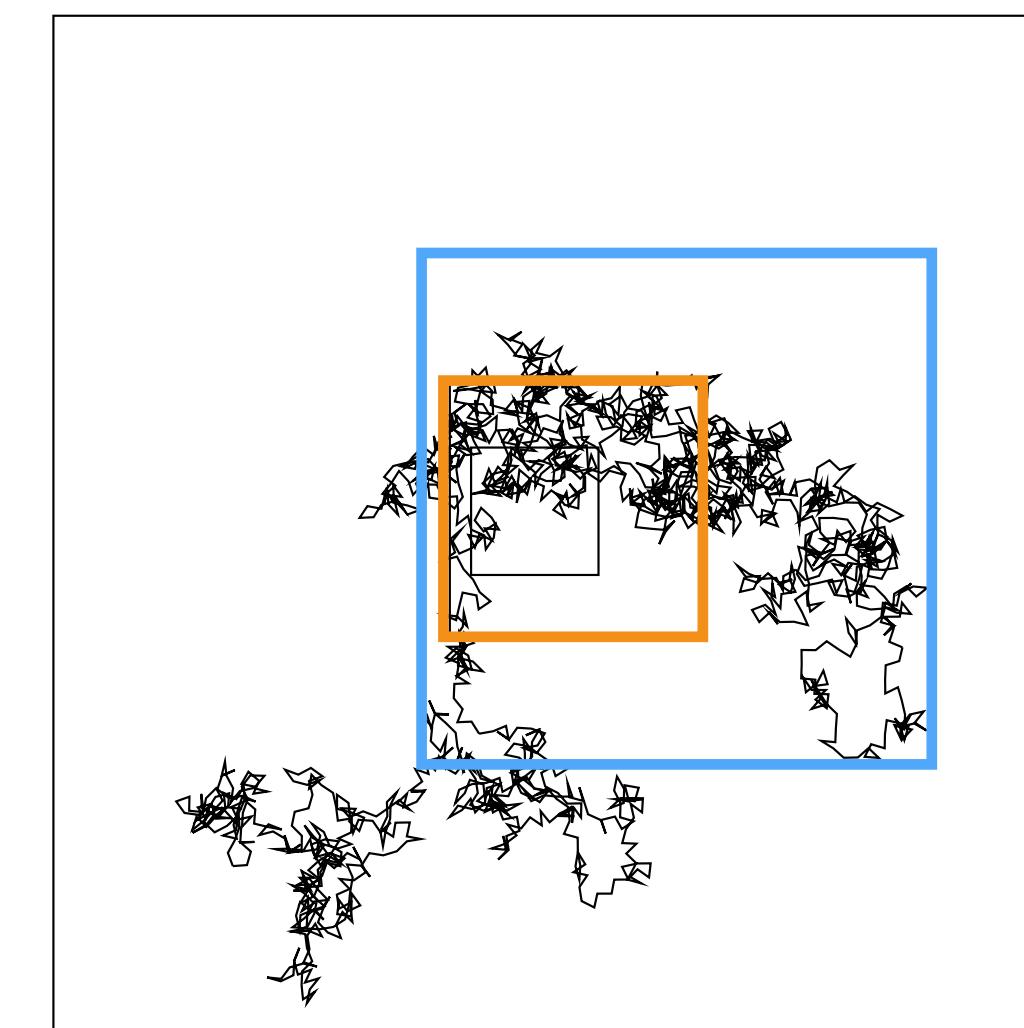
Scale invariance of random walks



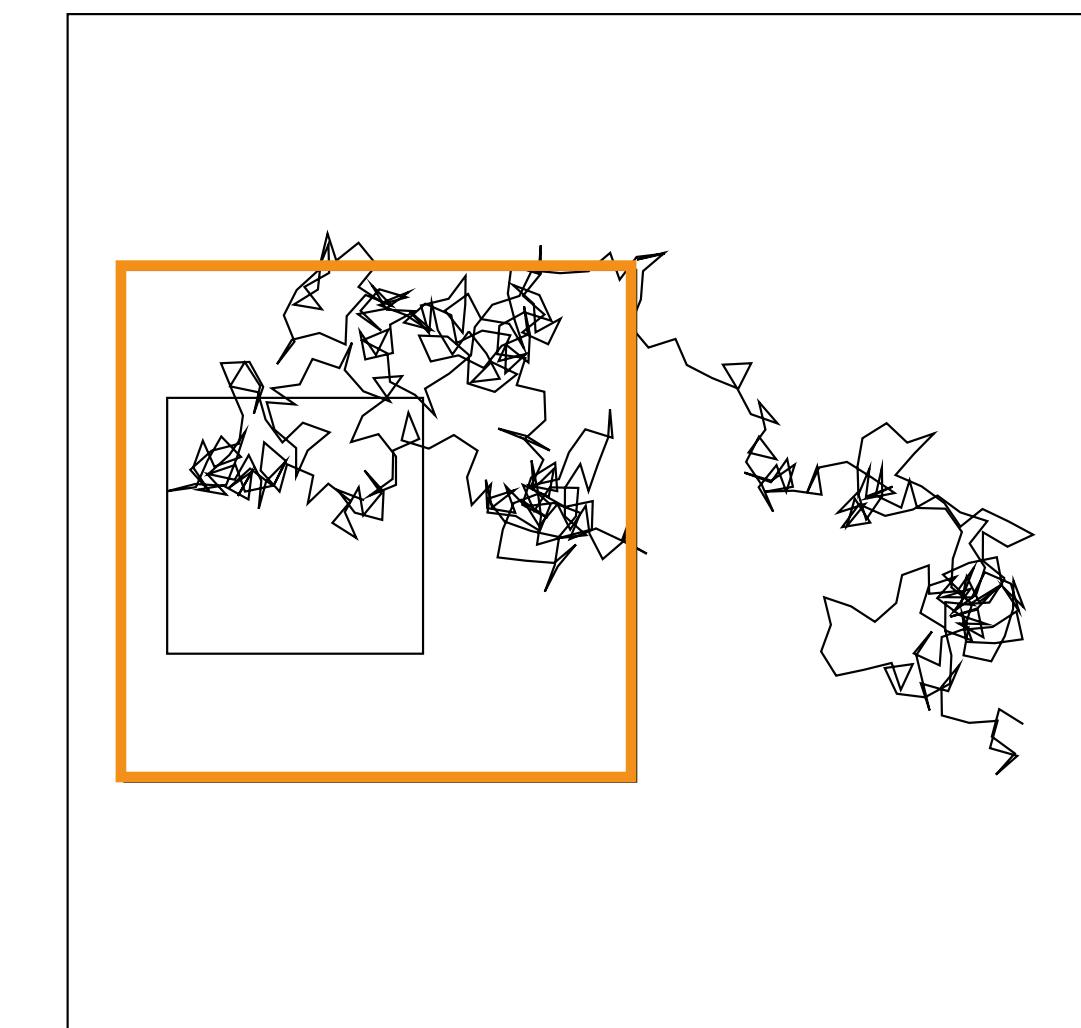
32 000 steps



First 8 000 steps



First 2 000 steps



First 500 steps

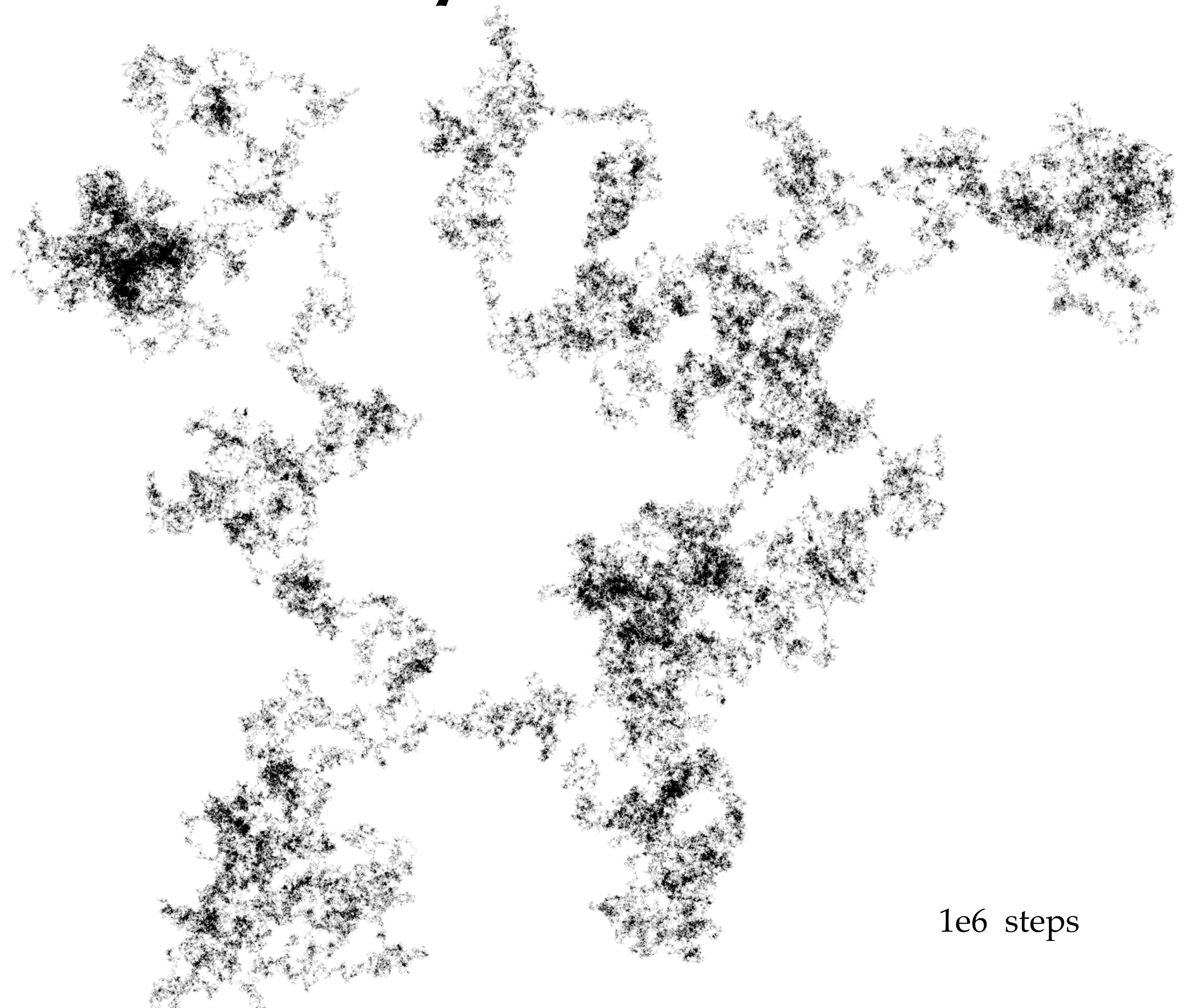
Random walks form a jagged,
fractal-like pattern...
which looks the same when rescaled.



scale invariant

Universality

Play with the simulations [here](#)



1e6 steps

Universality

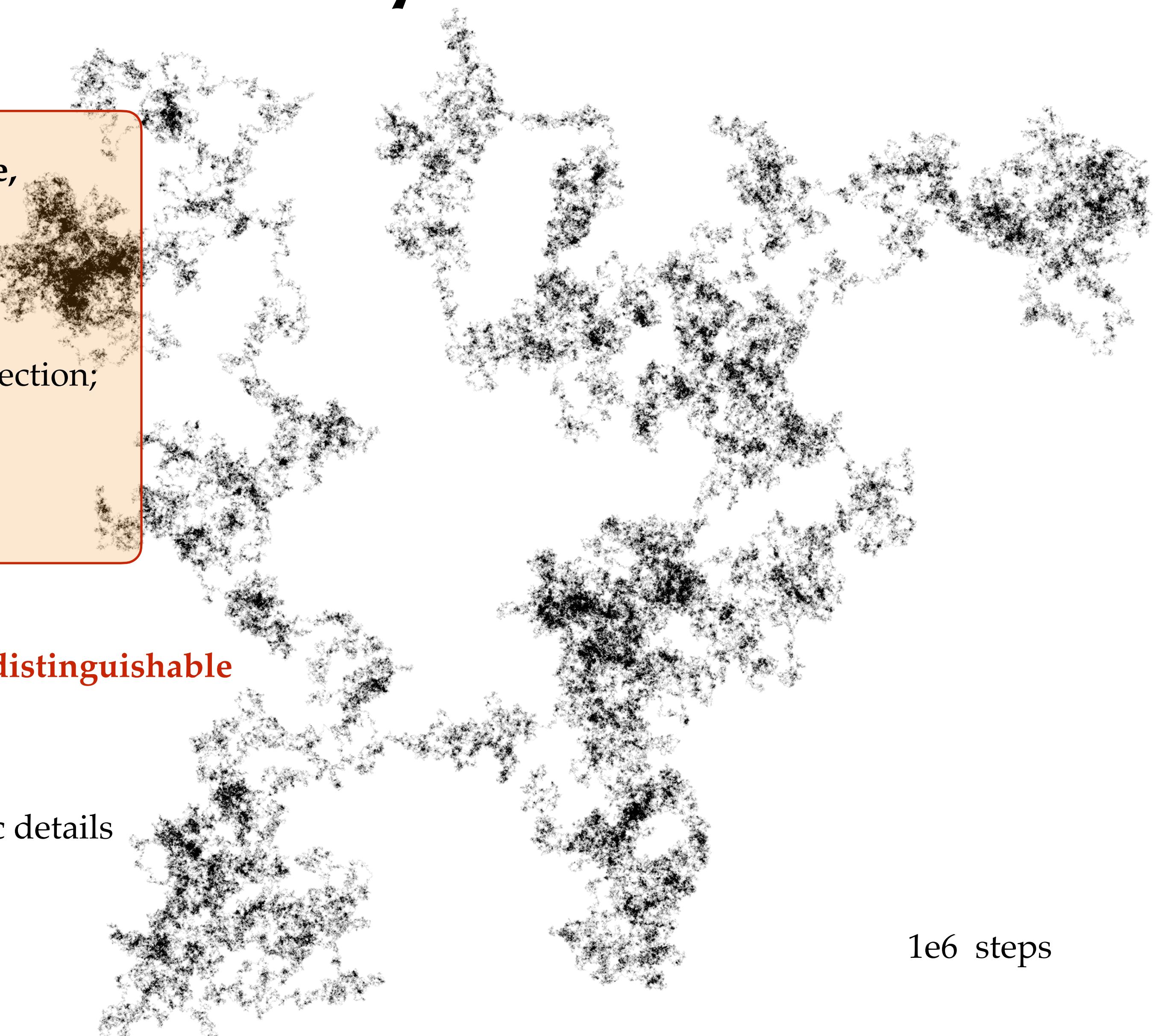
Play with the simulations [here](#)

On scales where the individual steps are not distinguishable,
all these random walks look the same:

- RW on a 2D grid;
- 2D RW with fixed length, but uniformly distributed direction;
- 2D RW uniformly distributed direction,
and exponentially jump lengths;

After a few steps, all of these random walks are **statistically indistinguishable**

Universality: independence of the behavior on the microscopic details



1e6 steps

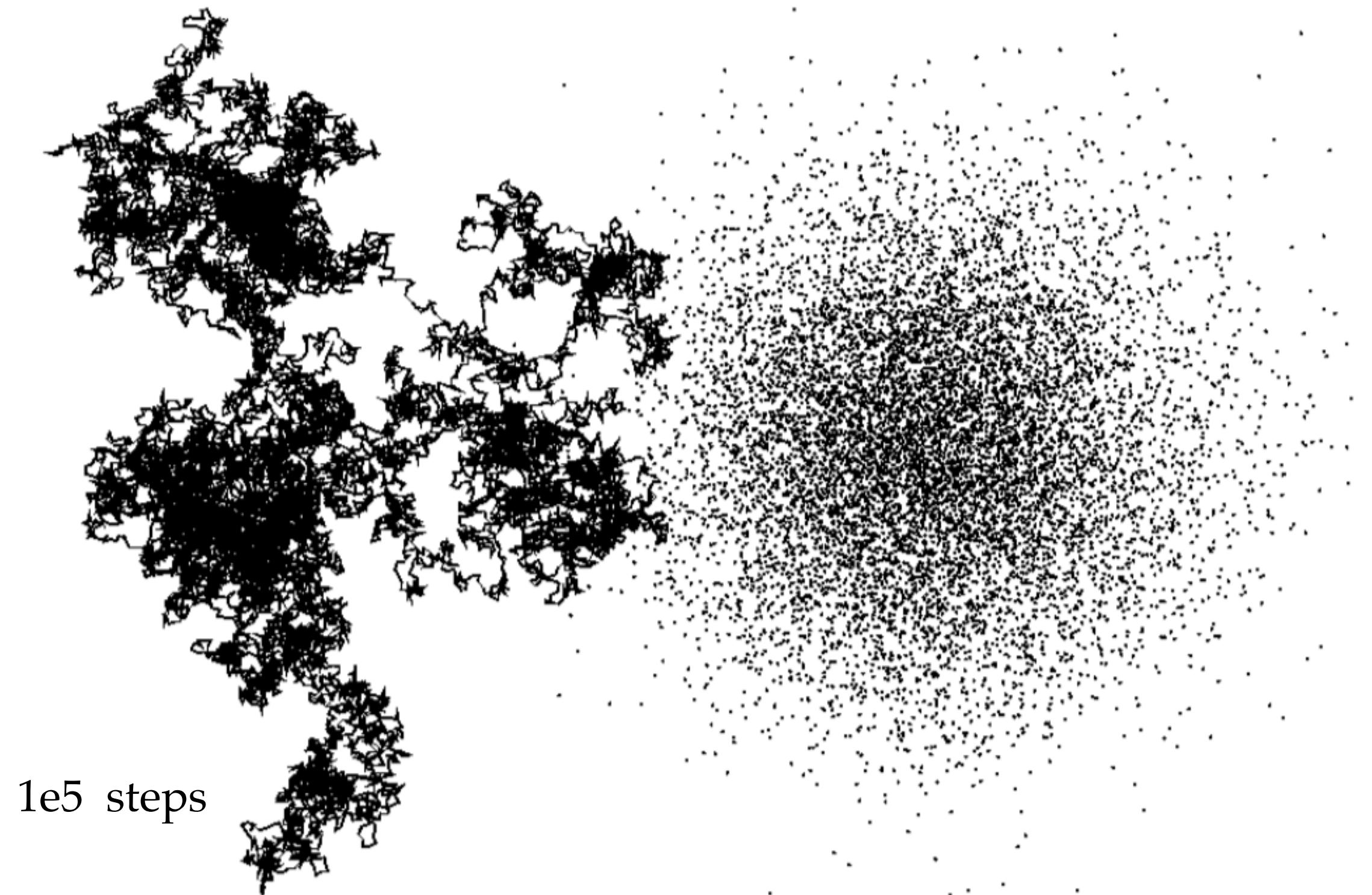
Emergent behavior

Simple laws/behaviors that emerge from underlying complexity

Observe on the simulations [here](#)

Two types of emergent behavior:

An **individual random walk**,
after a large number of steps,
becomes *fractal* or **scale invariant**



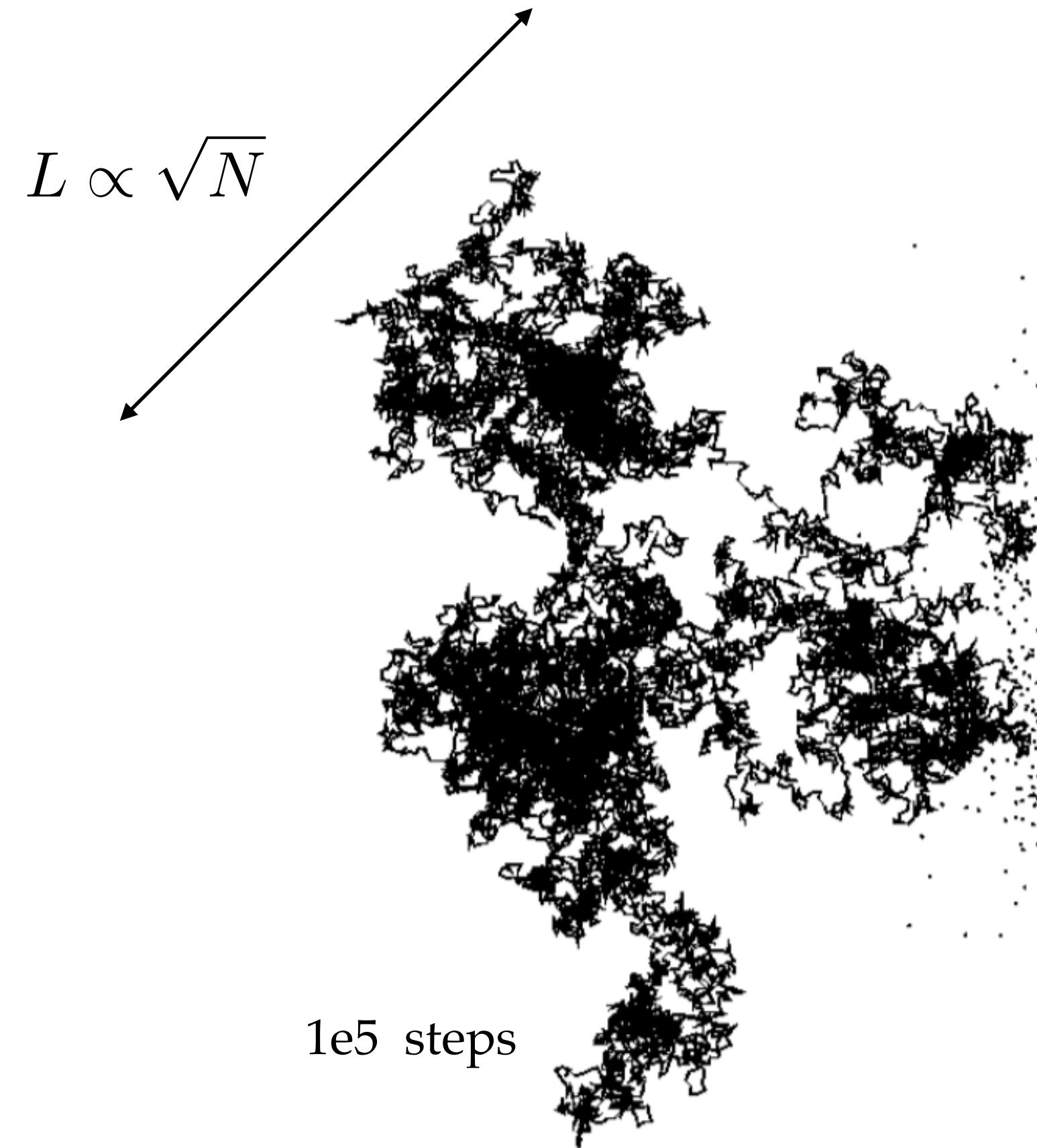
The **endpoint** of the random walk
has a **probability distribution**
that obeys a **simple continuum law**,
the diffusion equation

Both of these behaviors are **independent of the microscopic details** of the walk: they are **universal**.

Extension of a random walk

Simple laws/behaviors that emerge from underlying complexity

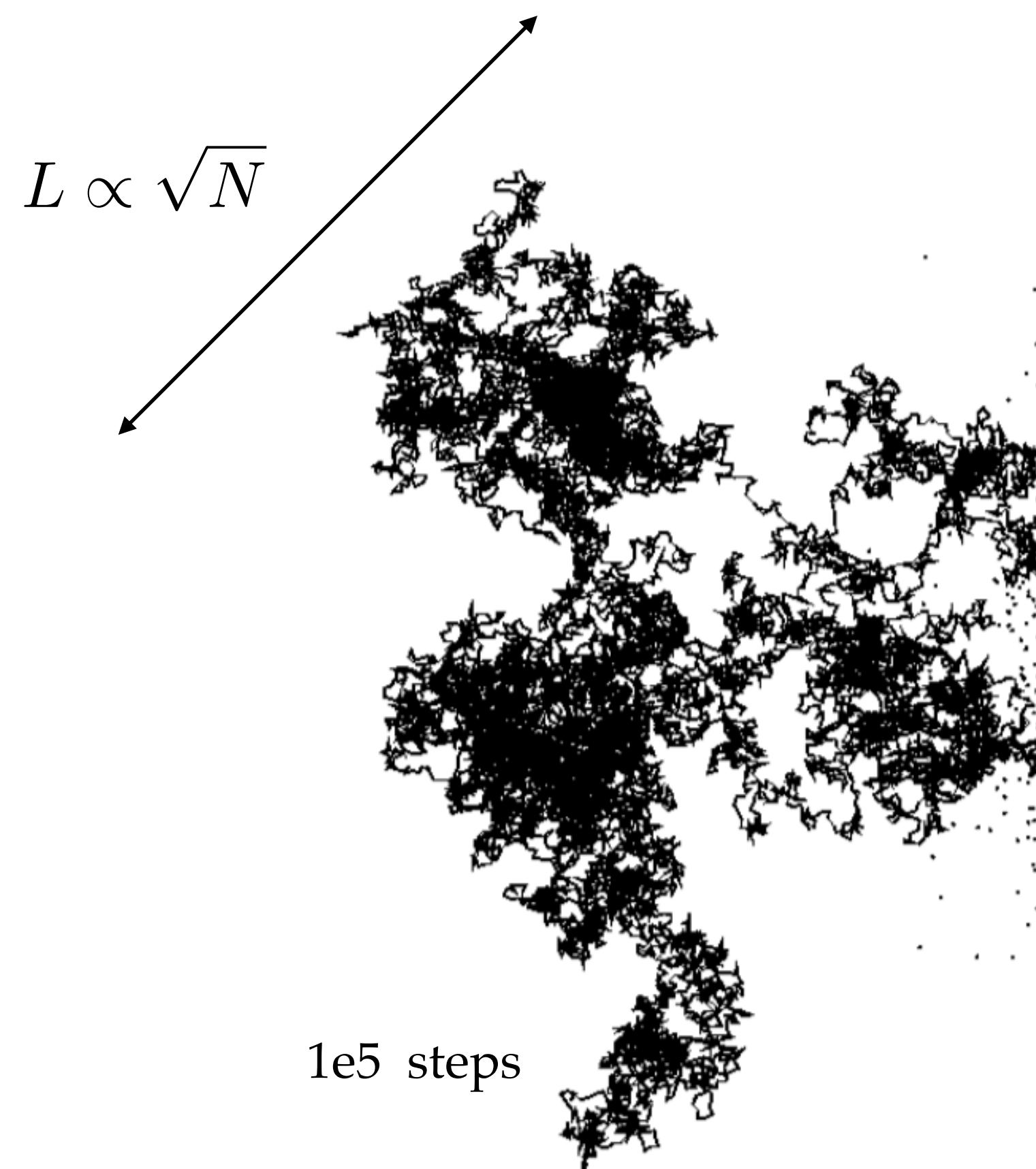
“Spatial extension” of the walk:



Extension of a random walk

Simple laws/behaviors that emerge from underlying complexity

“Spatial extension” of the walk:



Independent and Identically Distributed random variables (or, i.i.d. random variables):

For any variable X_N defined as the sum of N independently and identically distributed (i.i.d.) random variables r_i for which σ_{r_i} is finite, we have that:

- the mean of X_N is equal to N times the mean of r_i :

$$\langle X_N \rangle = N \langle r_i \rangle ; \quad (5.34)$$

- the variance of X_N is equal to N times the variance of r_i :

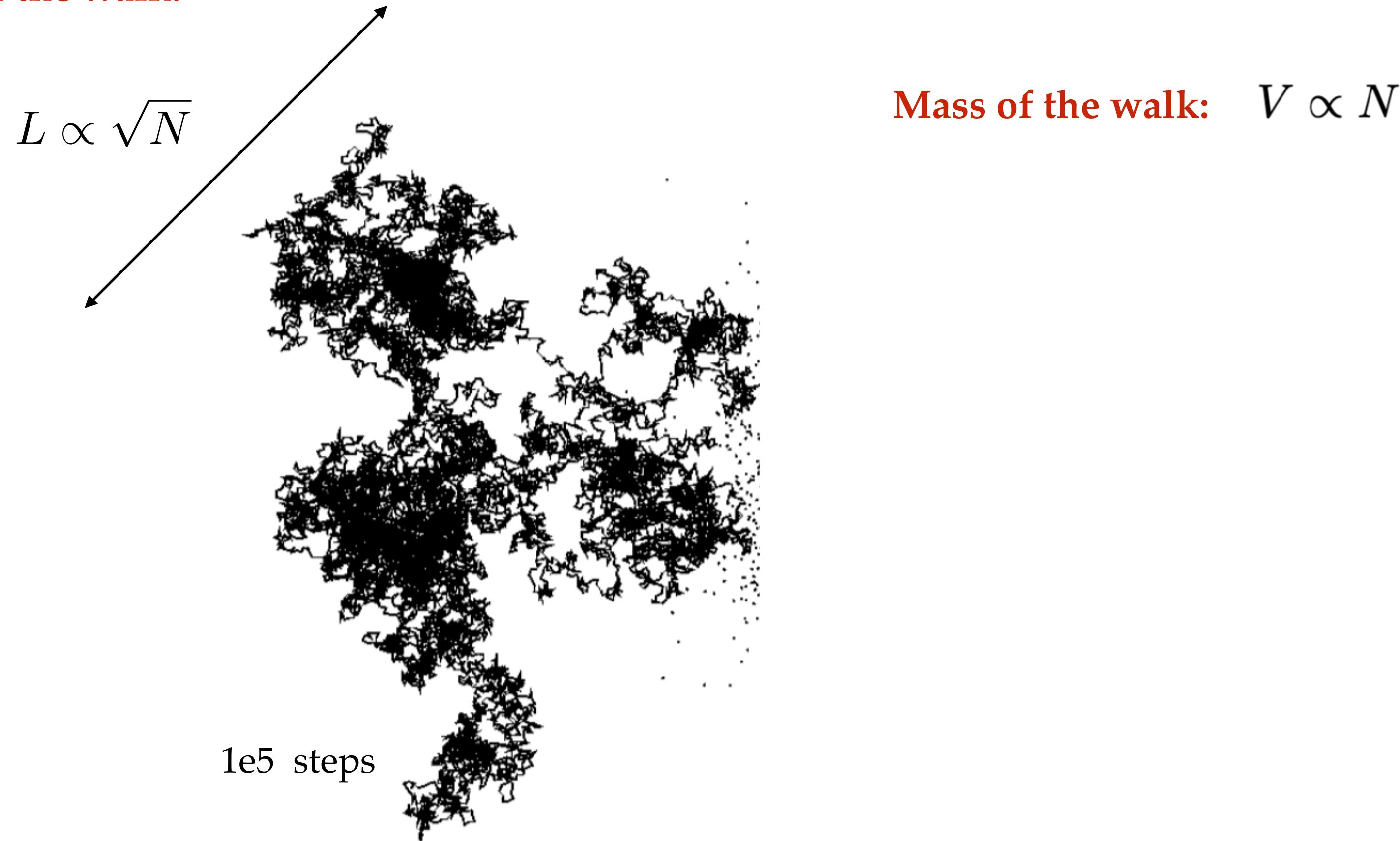
$$\sigma_{X_N}^2 = N \sigma_{r_i}^2 = N \left(\langle r_i^2 \rangle - \langle r_i \rangle^2 \right). \quad (5.35)$$

As a consequence, the standard deviation of the final positions of a 1-dimensional random walk defined by X_N always grows as \sqrt{N} , as long as σ_{r_i} is finite (no matter the detailed shape of the distribution of r_i).

Fractal dimension

Simple laws/behaviors that emerge from underlying complexity

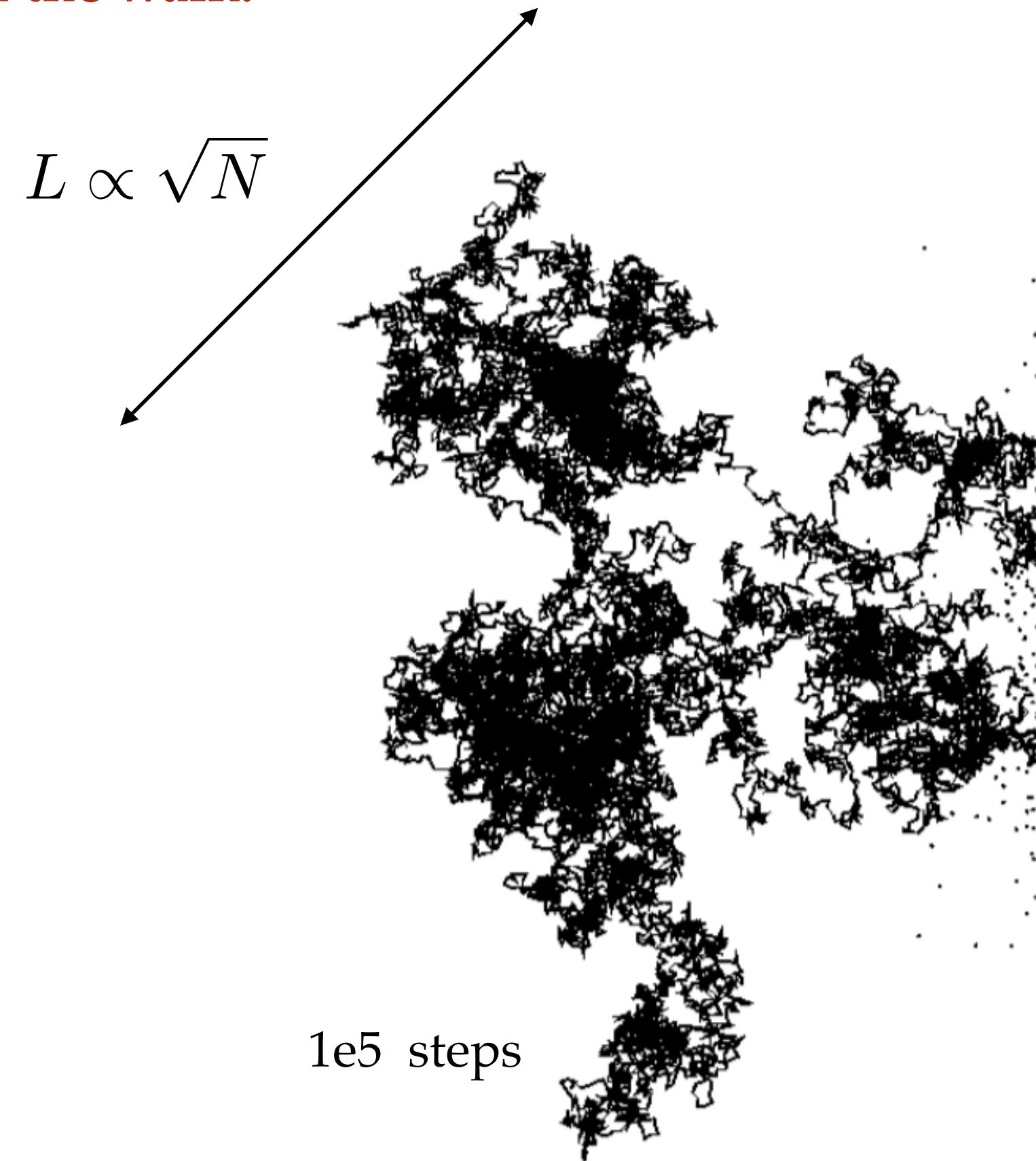
“Spatial extension” of the walk:



Fractal dimension

Simple laws/behaviors that emerge from underlying complexity

“Spatial extension” of the walk:



Mass of the walk: $M \propto N$

$$L \propto \sqrt{N} \quad \longrightarrow \quad L \propto \sqrt{M}$$

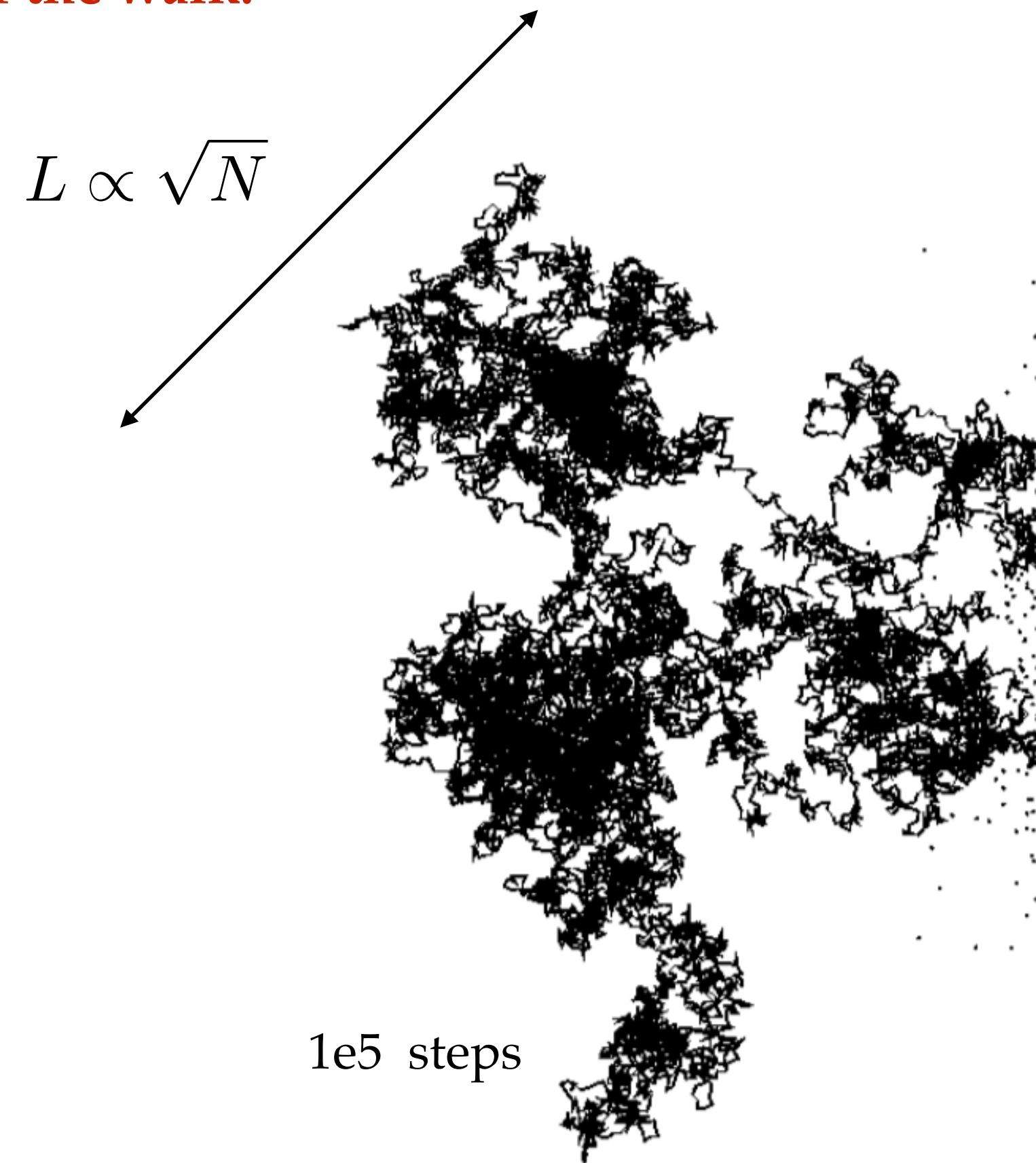
$$L_1 \quad \longrightarrow \quad L_2 = 2 L_1$$

$$M_1 \quad \longrightarrow \quad M_2 = 2^{d_F} M_1$$

Fractal dimension

Simple laws/behaviors that emerge from underlying complexity

“Spatial extension” of the walk:



Mass of the walk: $M \propto N$

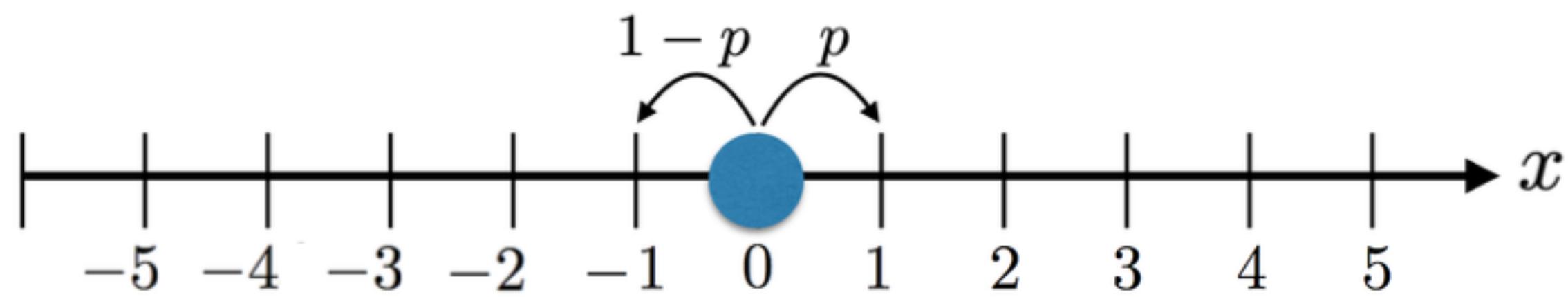
$$L \propto \sqrt{N} \quad \longrightarrow \quad L \propto \sqrt{M}$$

$$L_1 \quad \longrightarrow \quad L_2 = 2 L_1$$

$$M_1 \quad \longrightarrow \quad M_2 = 2^{d_F} M_1$$

$$d_F = 2$$

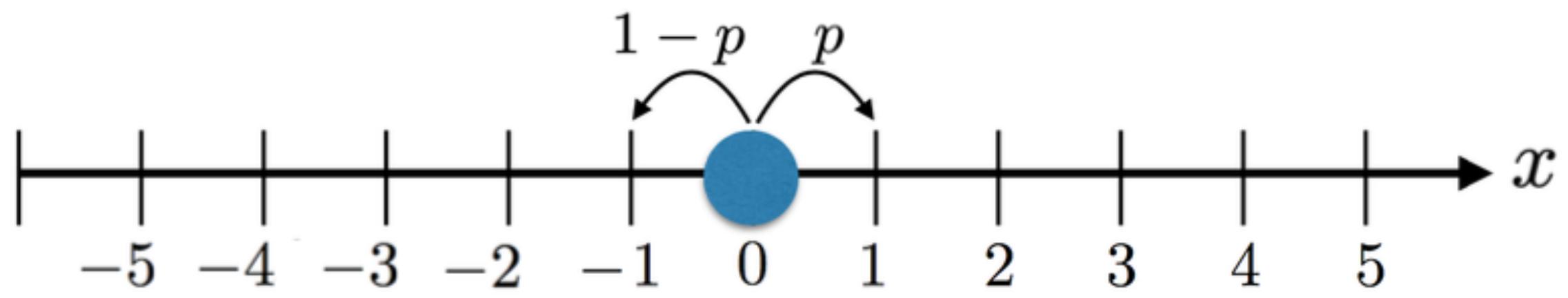
Master equation and Fokker-Planck equation



$P(x, N)$ = probability to find the walker in position x after N jumps.

Discrete Master equation:

Master equation and Fokker-Planck equation



Small jumps ℓ in small time intervals dt : $t = N dt$

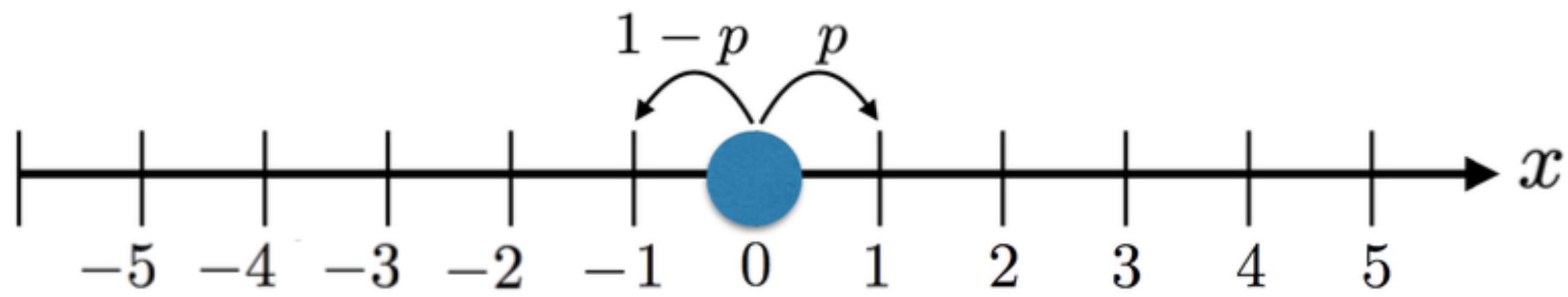
$$P(x, t + dt) = \underbrace{P(x - \ell, t) \times p}_{(1)} + \underbrace{P(x + \ell, t) \times (1 - p)}_{(2)}$$

$P(x, N)$ = probability to find the walker in position x after N jumps.

Discrete Master equation:

$$P(x, N + 1) = \underbrace{P(x - 1, N) \times p}_{(1)} + \underbrace{P(x + 1, N) \times (1 - p)}_{(2)},$$

Master equation and Fokker-Planck equation



Small jumps ℓ in small time intervals dt : $t = N dt$

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Taylor expansions:

- 1rst order in dt
- 2nd order in ℓ

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Discrete Master equation:

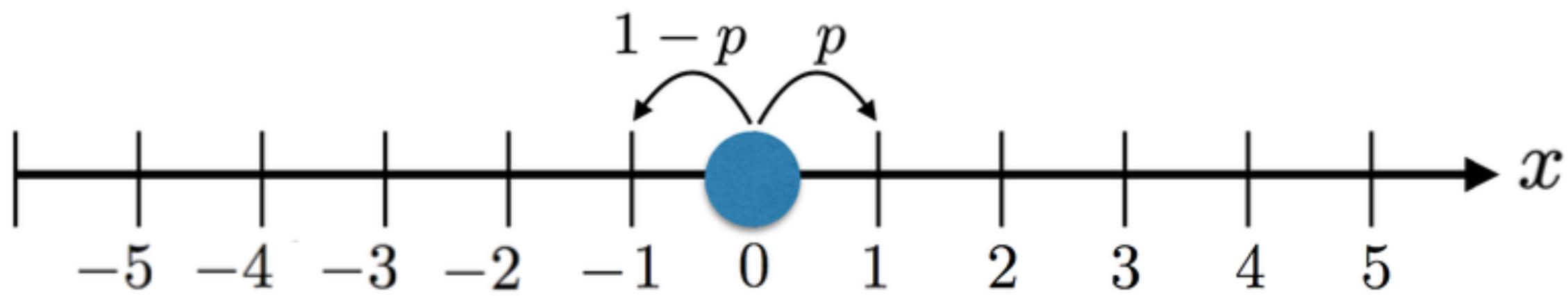
$$P(x, N + 1) = \underbrace{P(x - 1, N) \times p}_{(1)} + \underbrace{P(x + 1, N) \times (1 - p)}_{(2)},$$

☞ We recall that the Taylor expansion of a function $f(x)$ (of a single variable x) that is infinitely differentiable at a point x_0 is given by the infinite sum:

$$f(x_0 + h) = f(x_0) + h f'(x_0) + \frac{h^2}{2} f''(x_0) + \frac{h^3}{3!} f'''(x_0) + \dots \quad (5.42)$$

For very small values of h , one may then truncate this sum at a chosen order.

Master equation and Fokker-Planck equation



Small jumps ℓ in small time intervals dt : $t = N dt$

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For very small values of h , one may then truncate this sum at a chosen order.

$$\frac{\partial P}{\partial t}(x, t) dt + o(dt) = (1 - 2p) \ell \frac{\partial P}{\partial x}(x, t) + \frac{\ell^2}{2} \frac{\partial^2 P}{\partial x^2}(x, t) + o(\ell^2).$$

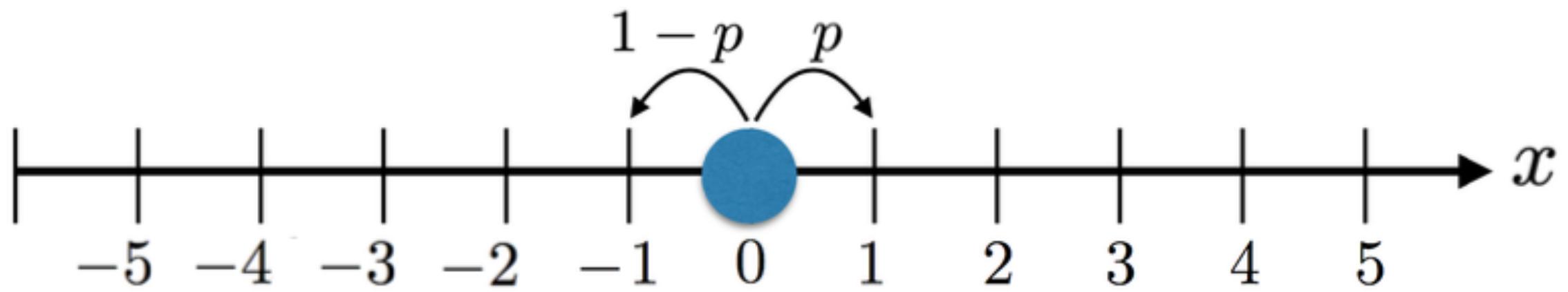
Fokker-Planck equation:

$$\frac{\partial P}{\partial t}(x, t) = -v \frac{\partial P}{\partial x}(x, t) + D \frac{\partial^2 P}{\partial x^2}(x, t),$$

diffusion coefficient $D = \ell^2 / (2 dt)$

drift $v = (2p - 1) \ell / dt$

Master equation and Fokker-Planck equation



Small jumps ℓ in small time intervals dt : $t = N dt$

$$P(x, t + dt) = \underbrace{P(x - \ell, t) \times p}_{(1)} + \underbrace{P(x + \ell, t) \times (1 - p)}_{(2)}$$

Taylor expansions: — 1rst order in dt
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$P(x, N)$ = probability to find the walker in position x after N jumps

Discrete Master equation:

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Diffusion equation:

$$p = 1/2$$

$$\frac{\partial P}{\partial t}(x, t) = D \frac{\partial^2 P}{\partial x^2}(x, t)$$

diffusion coefficient $D = \ell^2 / (2 dt)$

Fokker-Planck equation:

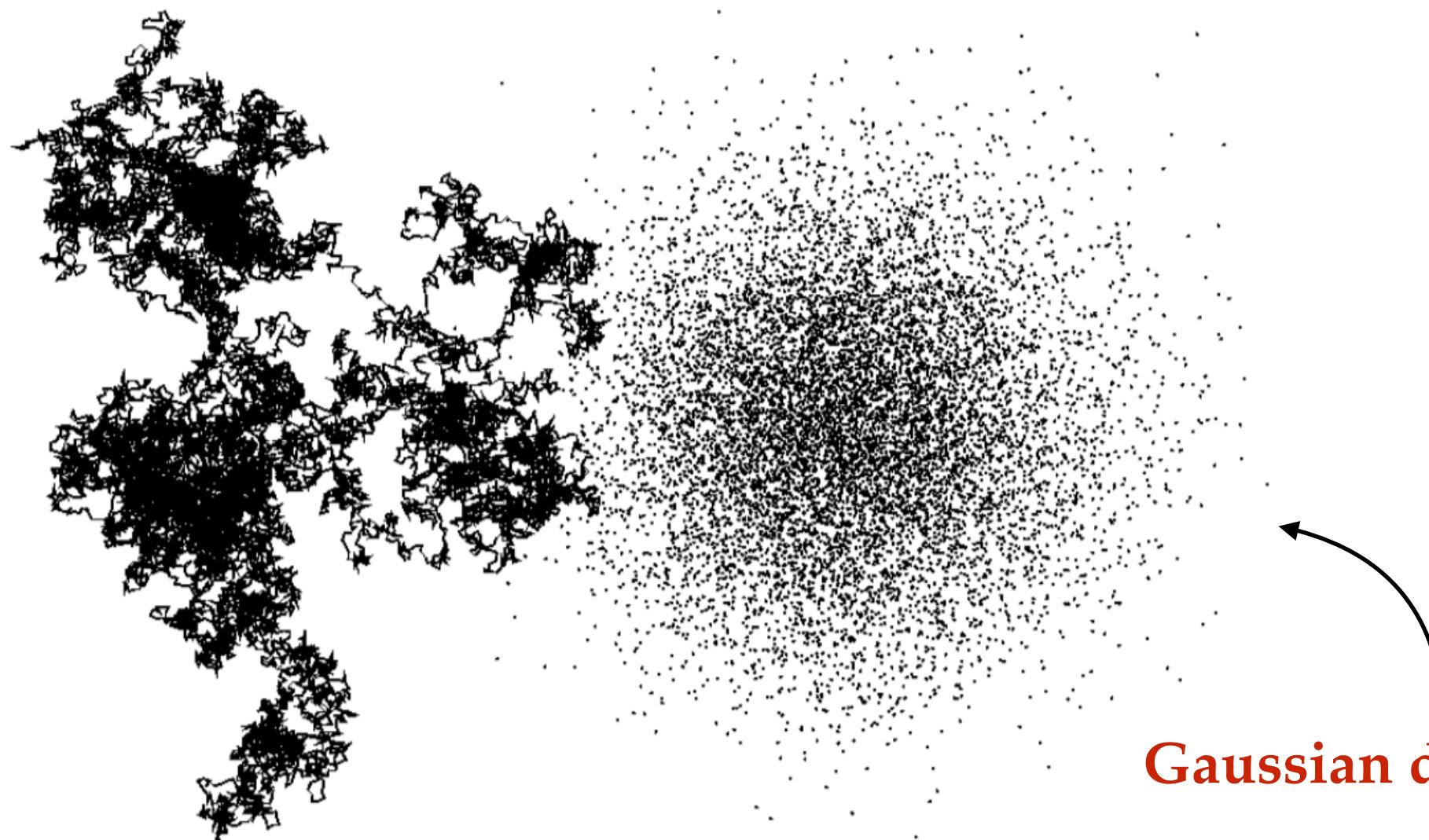
$$\frac{\partial P}{\partial t}(x, t) = -v \frac{\partial P}{\partial x}(x, t) + D \frac{\partial^2 P}{\partial x^2}(x, t),$$

diffusion coefficient $D = \ell^2/(2 dt)$

drift $v = (2p - 1) \ell/dt$

Solution:

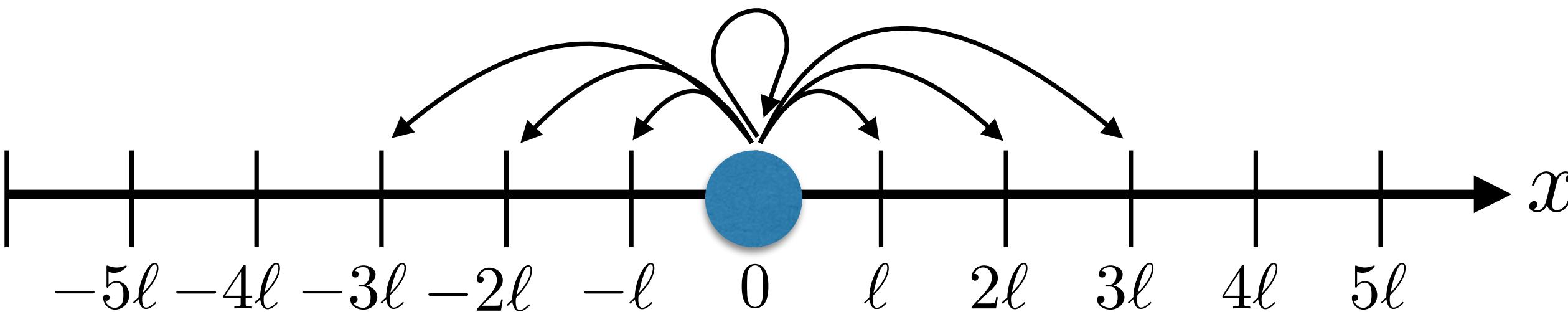
$$p(x, t) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x - \mu)^2}{2\sigma^2}\right) = \text{Gaussian } (\mu, \sigma^2) \quad \text{where } \mu = vt \\ \sigma^2 = 2Dt$$



The **endpoint** of the random walk
has a **probability distribution**
that obeys a **simple continuum law**,
the diffusion equation

Gaussian distribution of the endpoints

Fokker-Planck equation



In general the probability distribution Π of jump lengths s can depend on the particle position x $\Pi(s|x)$

Generalized master equation:

$$p(x, t + \Delta t) = \sum_s \Pi(s|x - s)p(x - s, t)$$

Again Taylor expand the master equation above to derive the Fokker-Planck equation:

$$\frac{\partial p(x, t)}{\partial t} = -\frac{\partial}{\partial x} \left[v(x)p(x, t) \right] + \frac{\partial^2}{\partial x^2} \left[D(x)p(x, t) \right]$$

drift velocity
(external fluid flow, external potential)

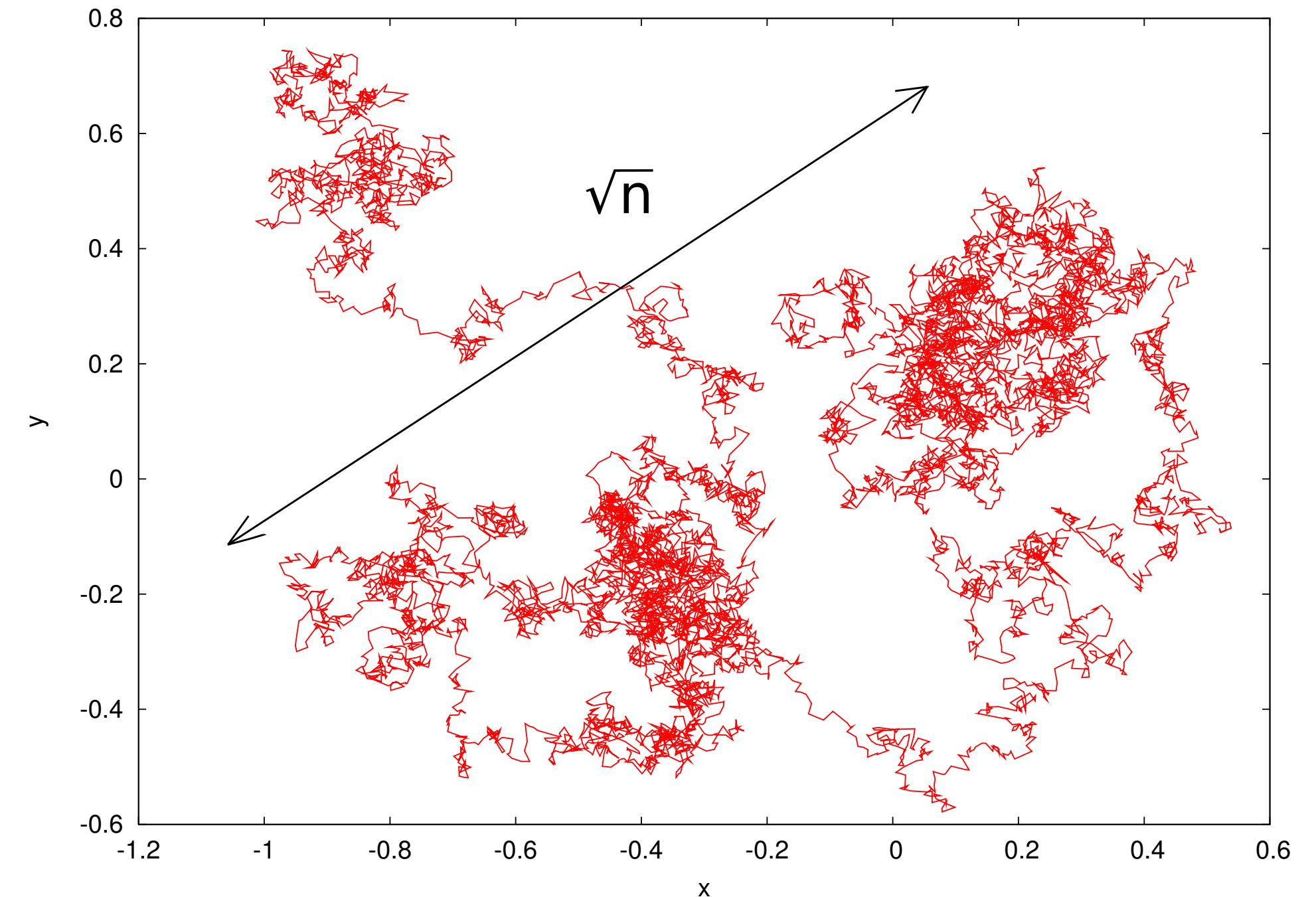
$$v(x) = \sum_s \frac{s}{\Delta t} \Pi(s|x) = \frac{\langle s(x) \rangle}{\Delta t}$$

diffusion coefficient
(e.g. position dependent temperature)

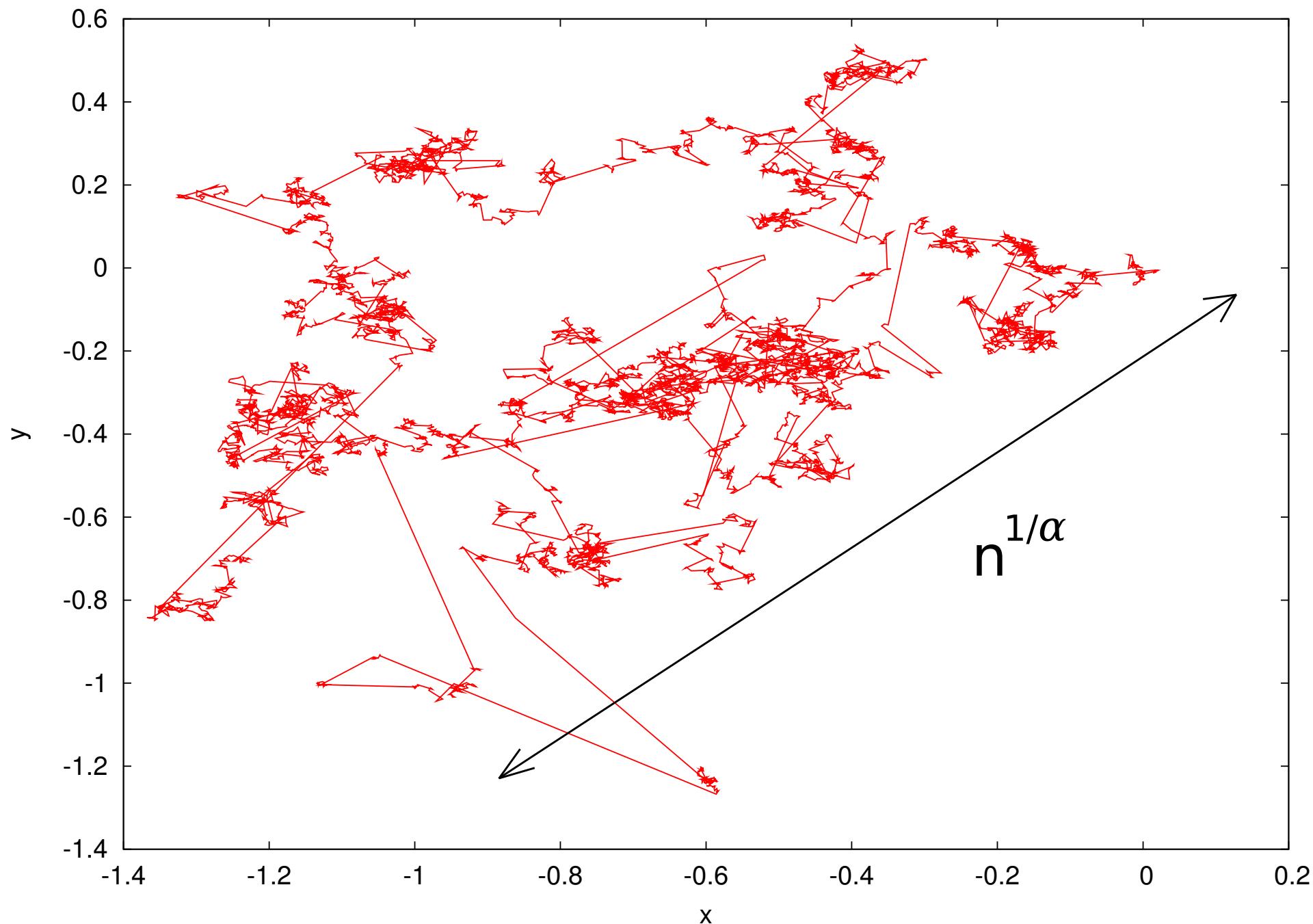
$$D(x) = \sum_s \frac{s^2}{2\Delta t} \Pi(s|x) = \frac{\langle s^2(x) \rangle}{2\Delta t}$$

Lévy flights

Jump length distribution: $\Pi(s) \sim \frac{c}{s^{\alpha+1}}$



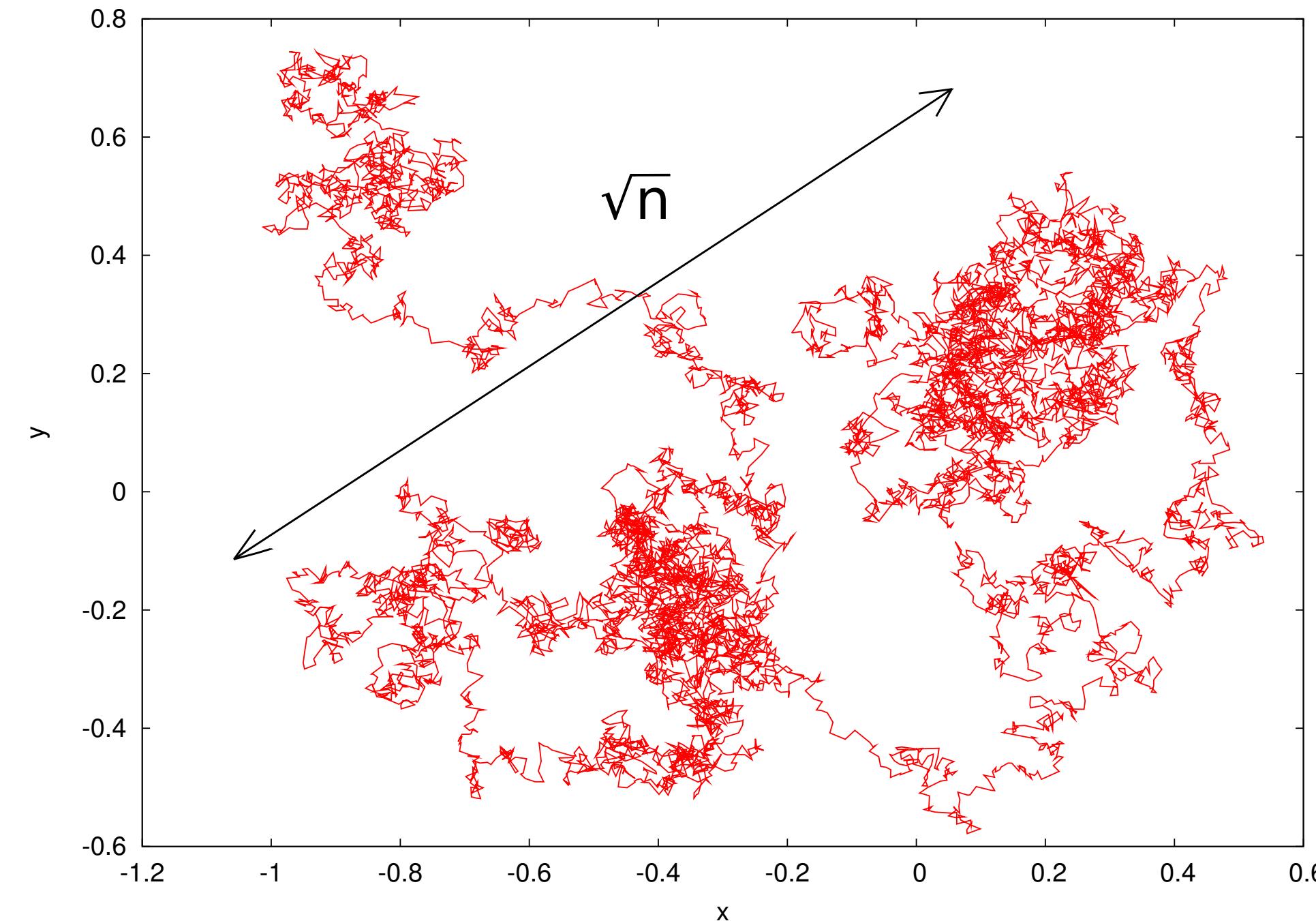
$$\alpha > 2$$



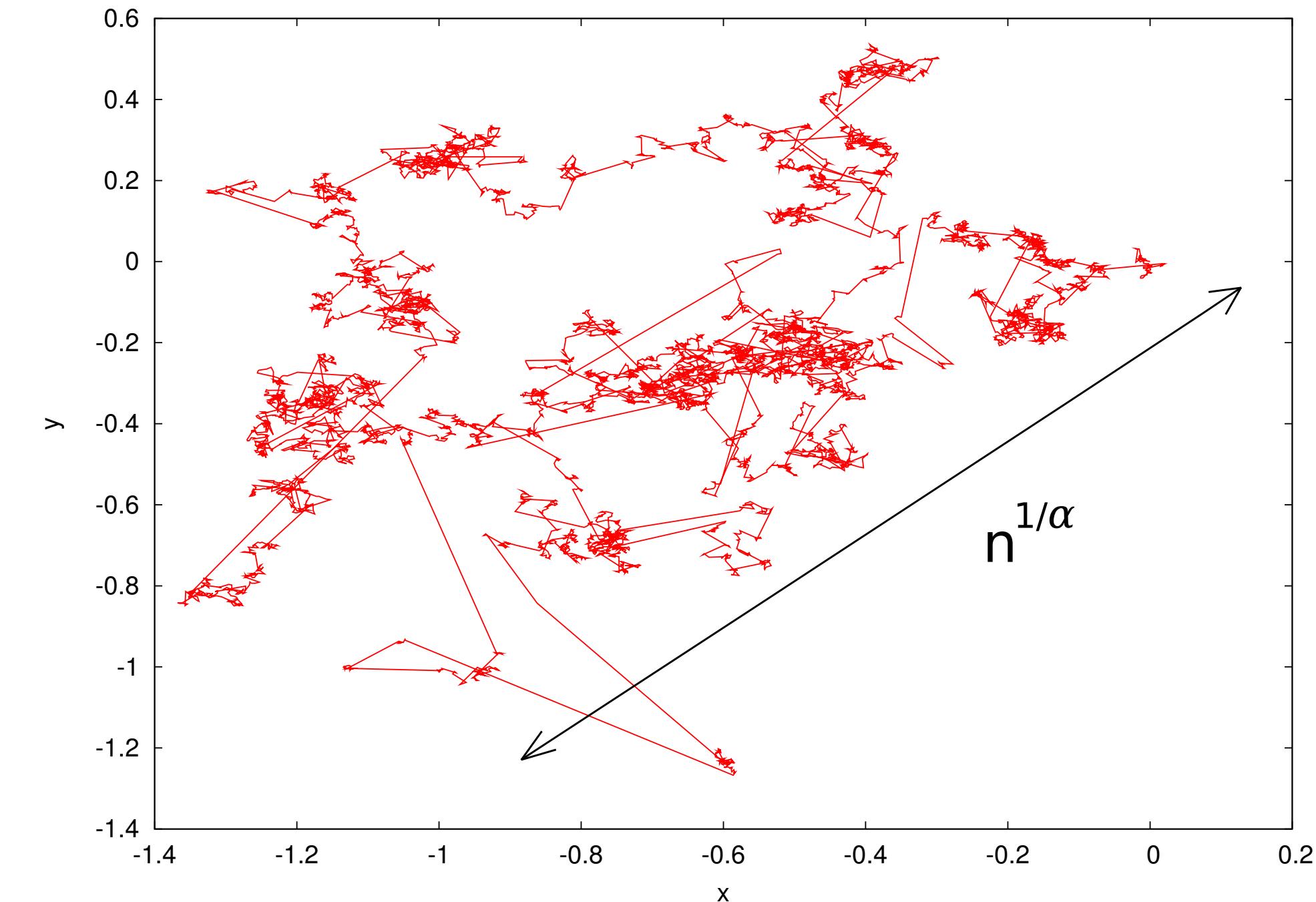
$$\alpha \leq 2$$

Lévy flights

Jump length distribution: $\Pi(s) \sim \frac{c}{s^{\alpha+1}}$

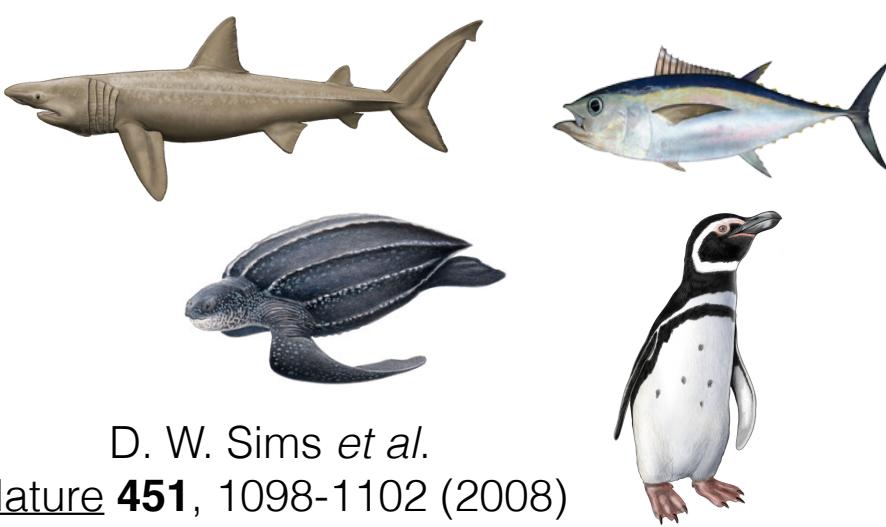


$$\alpha > 2$$



$$\alpha \leq 2$$

Spread faster than diffusion —> “Super-diffusion”

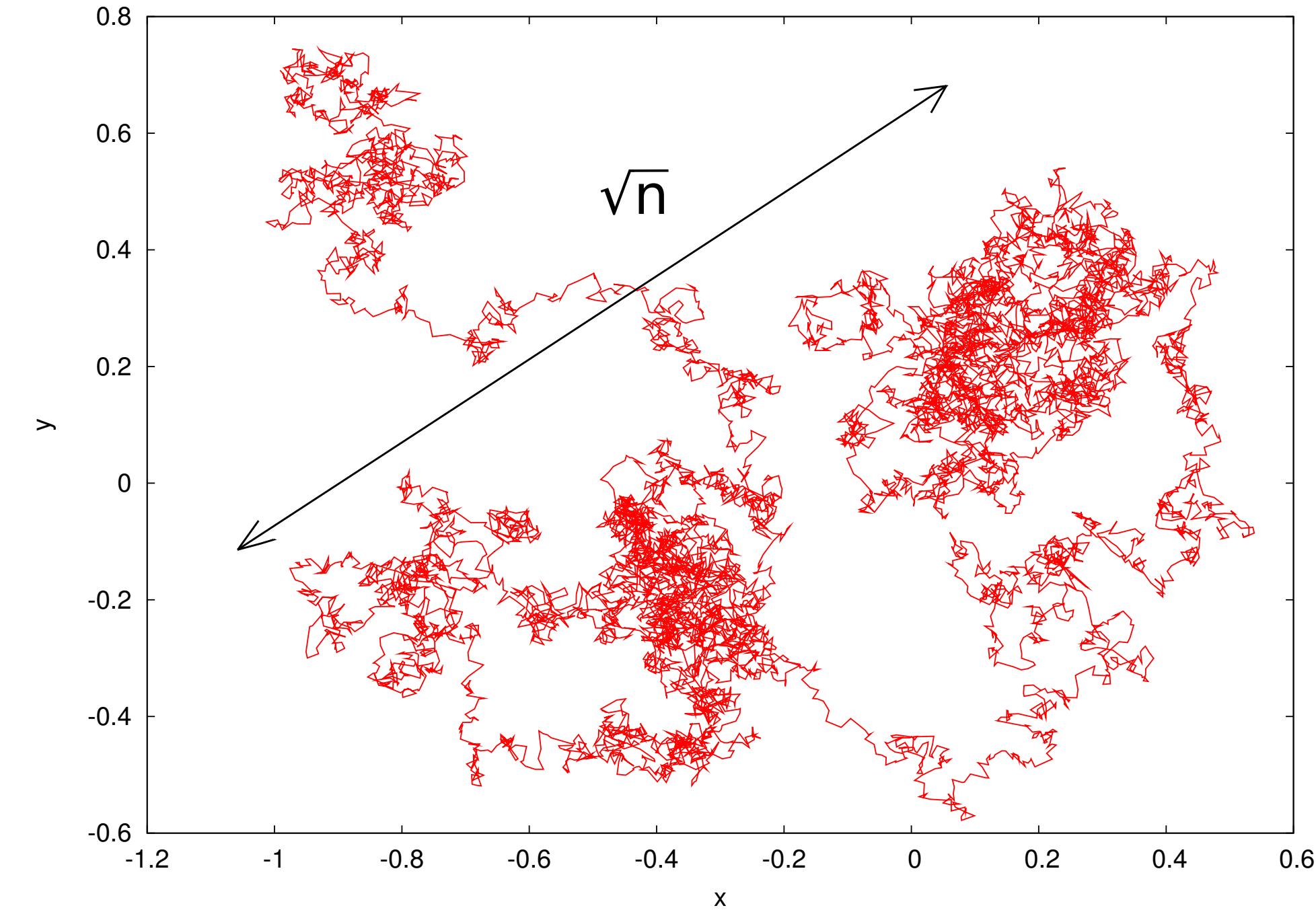


D. W. Sims *et al.*
Nature **451**, 1098-1102 (2008)

Argue that Lévy flights are a better search strategy
when prey is scarce

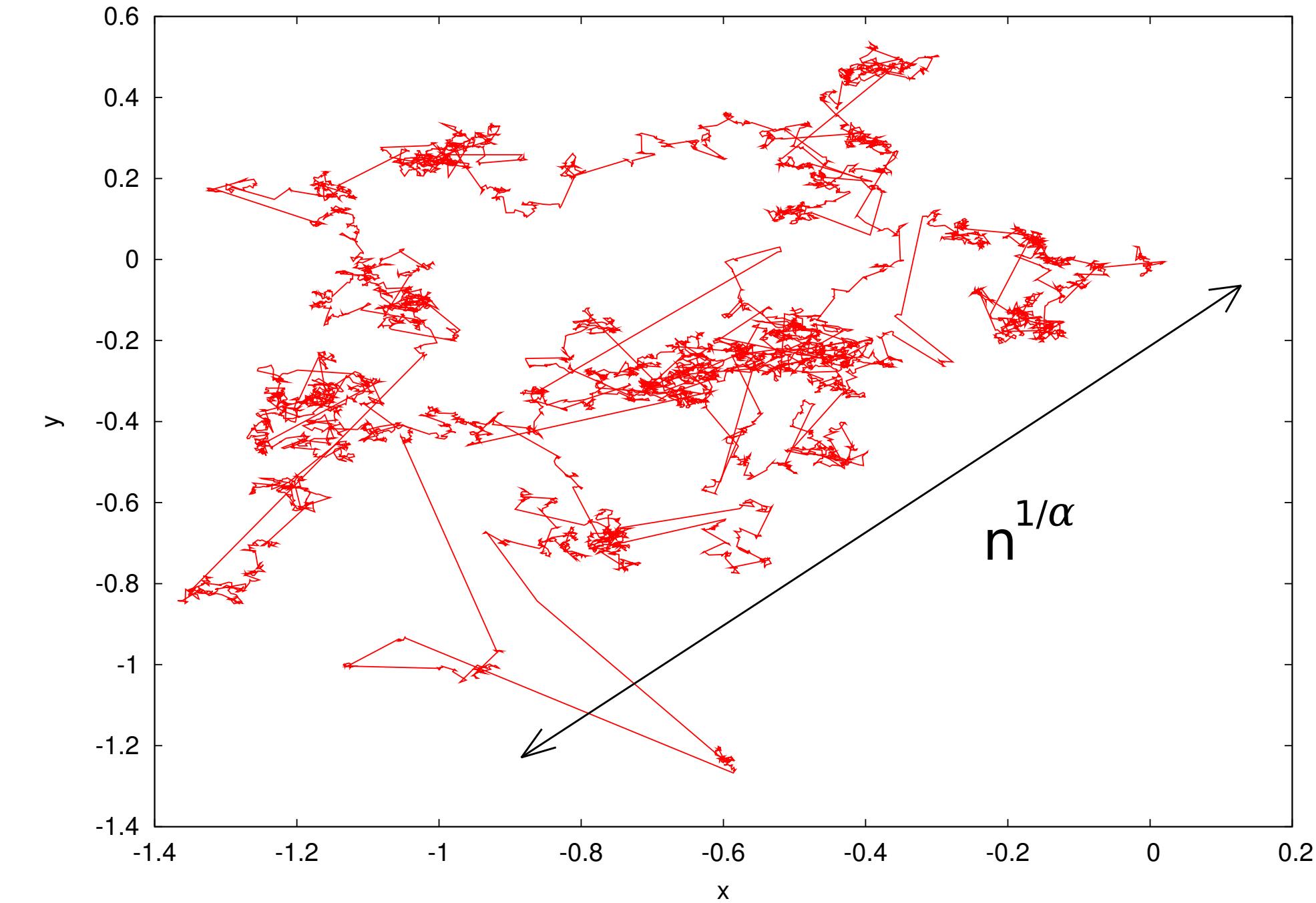
Lévy flights

Jump length distribution: $\Pi(s) \sim \frac{c}{s^{\alpha+1}}$



$$\alpha > 2$$

Mean and variance defined \rightarrow Brownian motion



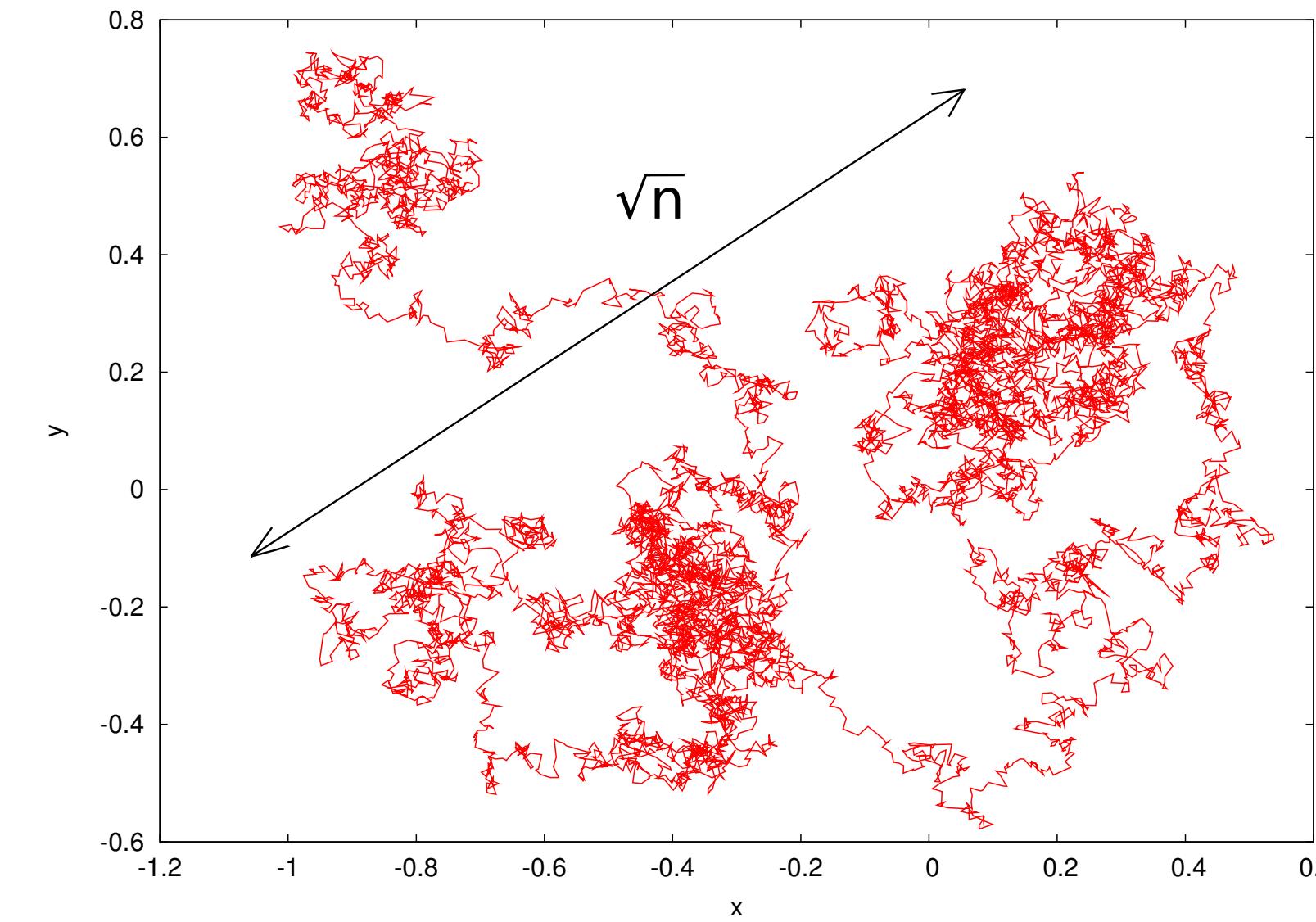
$$\alpha \leq 2$$

Variance not defined (infinite) \rightarrow some rare but very large jumps

\rightarrow Lévy Flights

Lévy flights

Jump length distribution: $\Pi(s) \sim \frac{c}{s^{\alpha+1}}$

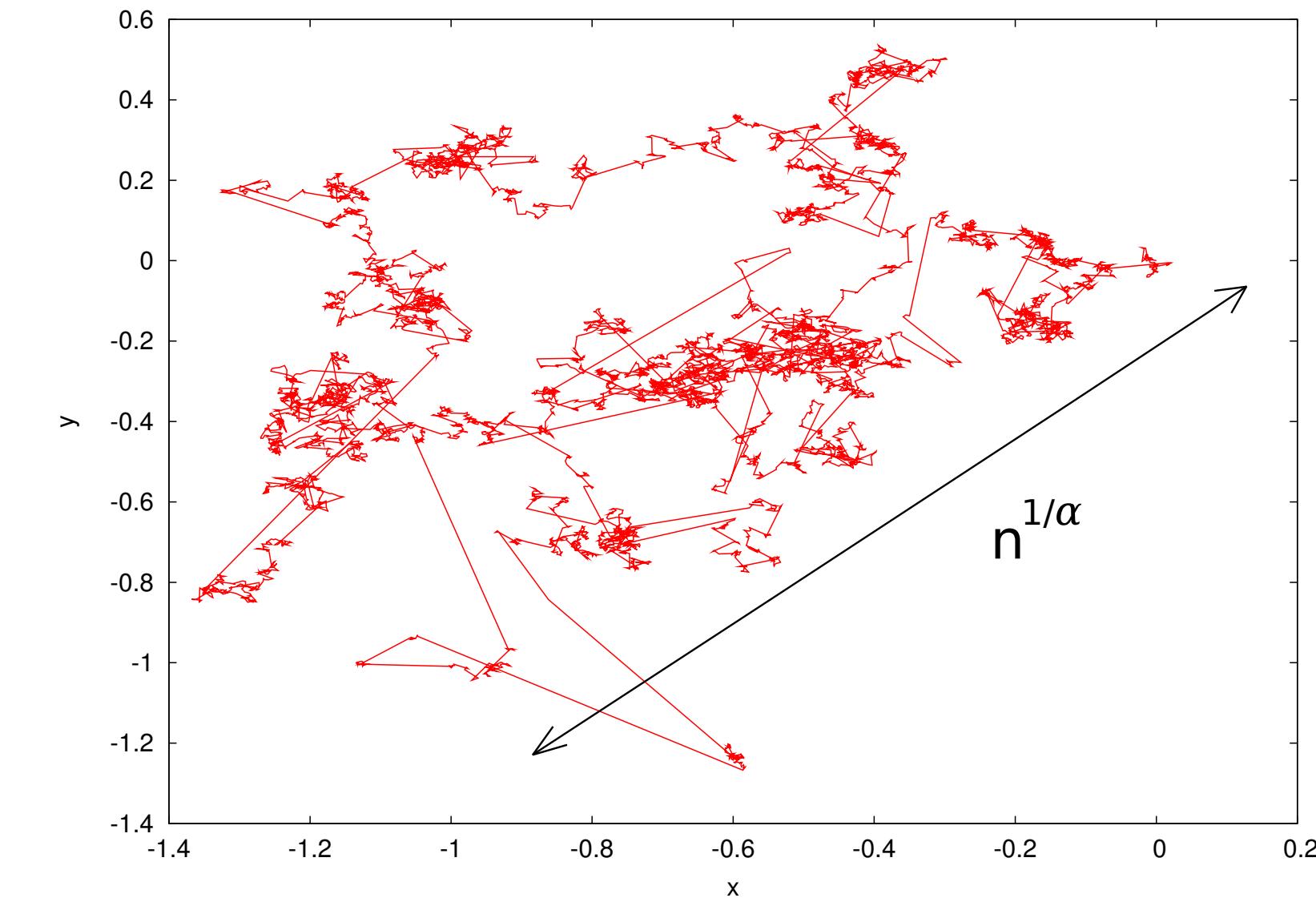


$$\alpha > 2$$

Mean and variance defined —> Brownian motion

Distribution of the walker position = Gaussian distribution (CLT)

Evolution described by: Diffusion Equation — Fokker-Planck Equation



$$\alpha \leq 2$$

Variance not defined (infinite) —> Lévy Flights

Distribution of the walker position

= Lévy stable distribution with same exponent α (generalized CLT)

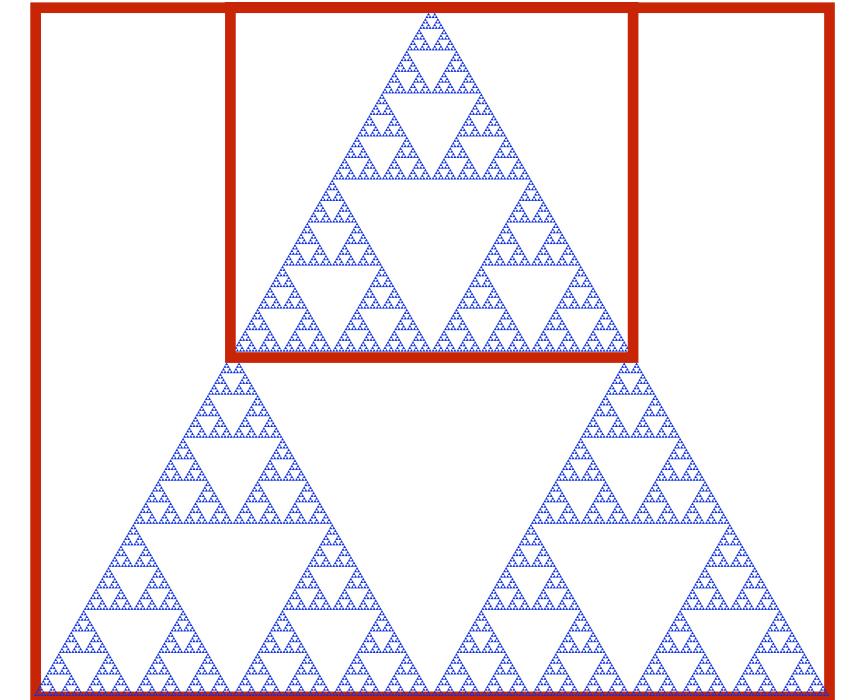
Evolution described by: fractional Fokker-Planck Equation

Self-Similarity and Scale Invariance

Self-similar object: is exactly or approximately similar to a part of itself

It is a typical properties of **fractals**:

Many real world elements are **statistically self-similar**:



Scale invariance: At any scale, there is a small piece of object that is similar to the whole.

