

Introduction to the mathematical modeling of
multi-scale phenomena
Diffusion

MCB/MATH 303

Brownian motion

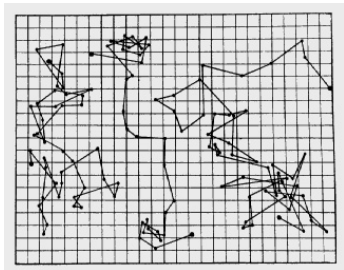
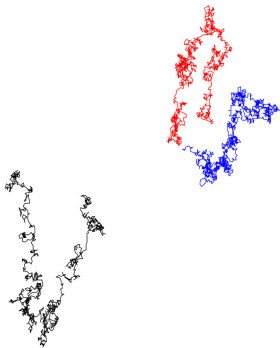


Image (public domain) from J. B. Perrin, *Mouvement Brownien et réalité moléculaire*, *Ann. de Chimie et de Physique* **18**, 5-114 (1909).

- **Brownian motion** (named after botanist Robert Brown) refers to the random motion of particles suspended in a fluid as they are “pushed around” by the smaller fluid molecules.
- This phenomenon may be modeled in terms of a **random walk**.
- The **diffusion MATLAB GUI** illustrates that, on average, a particle performing a random walk on the plane is, after time t , at a distance proportional to \sqrt{t} from the origin.

Random walk



Trajectories of three particles performing a 1000-step random walk on the plane.

- This immediately allows us to introduce a dimensionless quantity DT/L^2 , where T is a characteristic time, L a characteristic length, and D a **diffusion coefficient**.
- We recognize the **same dimensionless combination** that appears in the heat equation,

$$\frac{\partial u}{\partial t} = D \frac{\partial^2 u}{\partial x^2}.$$

One can in fact show that a **random walk** at the microscopic level can be described at the macroscopic level by a **diffusion equation**.

Reaction-diffusion equations

- The dynamics of macroscopic dynamics of quantities that **both diffuse and interact**, such as chemicals in a reaction, is typically described in terms of a **reaction-diffusion equation** of the form

$$\frac{\partial u}{\partial t} = D \frac{\partial^2 u}{\partial x^2} + F(u),$$

where F is a **reaction term**.

- Reaction-diffusion systems often lead to the formation of **Turing patterns**, such as hexagons or stripes.
- Such patterns, which have been reproduced in chemical reaction experiments, are also thought to be observed on **animal coats and fish skins**.



Microscopic versus macroscopic aspects of diffusion

- The **macroscopic description** of a **microscopic** phenomenon is typically obtained by taking **averages**. As a consequence, individual aspects are lost and replaced by quantities such as densities, which can be measured at the macroscopic level.
- Since a random walk at the microscopic level can be described at the macroscopic level by a diffusion equation, the same reasoning may be applied to describe any motion that involves a random walk, **regardless of the scale at which it happens**.
- In particular, **flagellated bacteria** such *Bacillus subtilis* swim in a series of runs and tumbles, which can be **modeled** by a random walk.

Motion of *B. subtilis* on agar

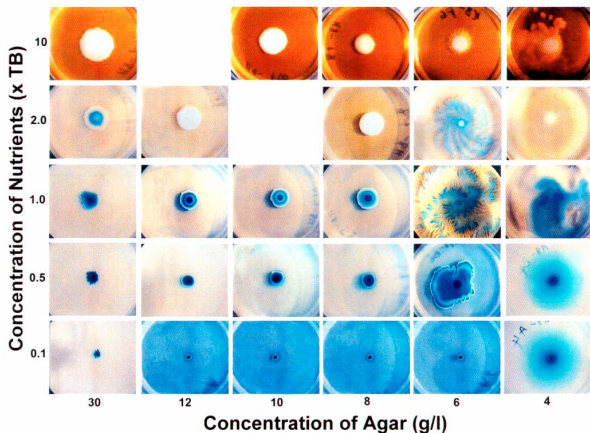
Bacillus subtilis is a flagellated rod-like bacterium

Movie (made in 2001 by then undergraduate student Cathy Ott in N. Mendelson's lab) showing *Bacillus subtilis* bacteria swimming on agar.

▶ Larger Movie

- Length: 2 to 3 μm .
- Diameter: $\sim 0.7 \mu\text{m}$.
- **Swimming speed:** about 10 times its length per second.
- It moves by a succession of **runs and tumbles**.

Colony forms for *B. subtilis*



N.H. Mendelson, and B. Salhi, *Patterns of reporter gene expression in the phase diagram of Bacillus subtilis colony forms*, J. Bacteriol. **178**, 1980-1989 (1996).

(Diameter of petri dishes: 6 cm)

Reaction-diffusion models

It is therefore not surprising that branched bacterial colony shapes may be captured by means of **reaction-diffusion models**.

$$\frac{\partial S}{\partial t} = D^S \nabla^2 S - \eta NS$$

S : density of nutrients
 N : density of bacteria

- These models often involve **nonlinear diffusion**,

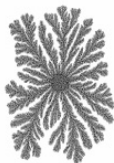
$$\frac{\partial N}{\partial t} = \nabla \cdot \left(D^N N^k \nabla N \right) + NS - \mu N$$

S. Kitsunezaki, J. Phys. Soc. Jpn. **66**, 1544-1550 (1997)

- Possibly with a **stochastic diffusion coefficient**:

$$\frac{\partial N}{\partial t} = \nabla \cdot \left(D^N (1 + \sigma) NS \nabla N \right) + NS$$

K. Kawasaki et al., J. Theor. Biol. **188**, 177-185 (1997)



Images from I. Golding et al., Physica **A 260**, 510-554 (1998).

Foraging behaviors

- It is natural to **push the analogy even further**, to describe the dynamics of “objects” of even bigger size, such as **animals foraging for food**.
- But of course, the scaling law $L^2 \propto T$ is **only valid** if the directions allowed for each step in the random walk are equally likely, and if steps are taken at regular intervals.
- In particular, **if the random walker stops** and stays put for a while, the average distance L after time T will grow more slowly than \sqrt{T} . This is called **subdiffusion**.
- Similarly, **if the walker takes unusually large steps**, L^2 will grow faster than T and **superdiffusion** will be observed.
- The above ideas have been applied in **ecology** to describe the **foraging behaviors** of animals.

Summary

- We started with **multi-scale aspects** of various physical or biological systems.
- We introduced the concept of **scales** and discussed examples of **scale-free** systems, such as **fractals** and **self-similar systems** found in nature.
- We then turned to a discussion of how **ideas of scales** appear in **models** that involve differential equations. This led us to **dimensional analysis** and **self-similar solutions** of partial differential equations.
- We studied **random walks** and used scalings to associate them with diffusion at the macroscopic level. This took us back to **patterns** via **reaction-diffusion equations**.
- Finally, we **extended** the concept of a random walk to phenomena occurring at the scales of a few microns (bacteria) and at our own scale (foraging animals).

Motion of B. subtilis on agar

Movie by Cathy Ott

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