

# The geometry of bacteria colonies

The microbiological background of motion, morphology diagram, Fisher equation. Self-affine surfaces, branching morphology and models for individual bacteria.

# Heuristic methods in physics

## “Traditional” approach

- You have a well defined question: how is acceleration ( $a$ ) related to the mass ( $m$ ) / How is pressure ( $p$ ) related to the volume ( $V$ ) in gas?
- You fix all the other parameters (keep the external conditions unchanged / molecule number, temperature, etc.) except those that are of interest:
  - You modulate one of them (e.g.  $m/V$ )
  - You measure the other ( $a/p$ )
  - Then you plot the results (and you have your answer)

## “Heuristic” approach

- You have an (often) less well-defined question about (often) more complex systems: How do bacteria colonies grow? How do fish or birds move?
- You assume which might be the relevant parameters (bacteria: nutriment, surface / fish: velocity, density)
- You assume a relation between them (creating a model)
- Study the behavior of the model (computational models!)
- Comparing (some aspects) with the original system
- But it still has:
  - descriptive
  - predictive power

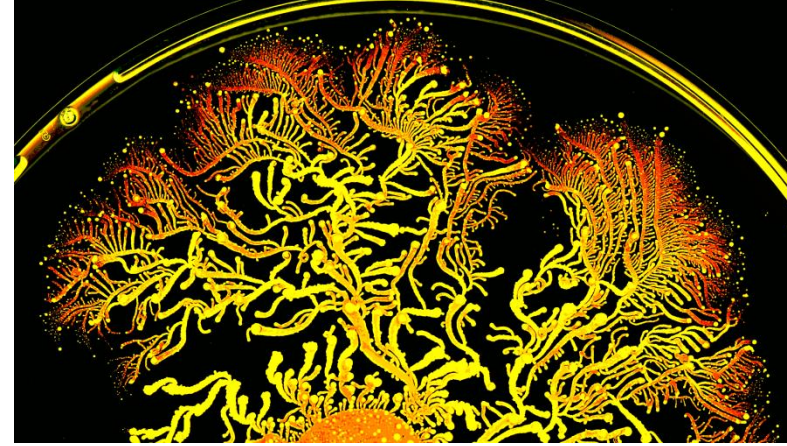
“Heuristics”: is a speculative, non-rigorous argument, that relies on an analogy or on intuition, that allows one to achieve a result or approximation to be checked later with more rigor; (Wikipedia)

# Why not the traditional way?

- These “more complex” systems are usually biological (even social) systems
- These systems are affected by...
  - Temperature
  - Humidity
  - Nutrients concentration
  - Group size
  - Group composition (age, gender, etc.)
  - The distribution of these
  - Inner social structure of the group
  - Transient inner states of the members (hunger, fear, etc.)
  - Reproduction stage of the members
  - Rain / sunshine
  - A boy throwing a ball into the flock (of birds)
  - The neighbor cat roaming around
  - A child crying for ice-cream nearby
  - Etc...
- It is hard to create “unchanged” conditions for measurements

# Why exactly bacteria colonies?

From a quantitative point of view



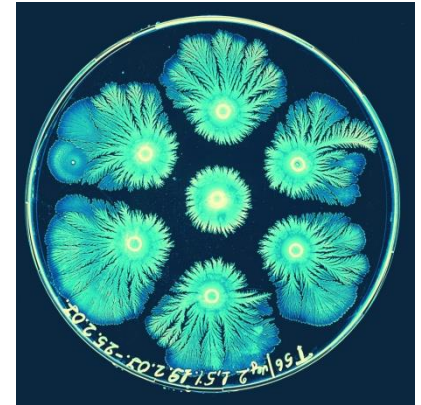
Colony of *Paenibacillus vortex* bacteria

- It is possible to “keep the environment unchanged”
- A system with interactions that are simple enough to be captured by quantitative models  
(The interaction rules are more or less understood)
- A system whose collective behavior can be explored with computational models  
(Theories can be modeled and tested via computer simulations)
- They can give an insight into the formation of self-organized biological structures

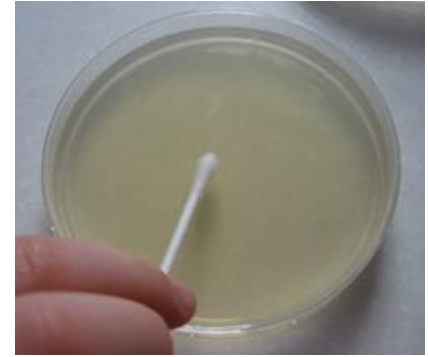
# Why exactly bacteria colonies?

From a biological point of view...

- Unicellular organisms
- Living in colonies
- They are easy to handle in experiments
- They produce various spatio-temporal patterns
- The patterns are often independent of the interaction details - “universality”
- Dependency on environmental conditions
- Experiments can be reproduced



# The set-up of the simplest experiments for colony formation



Bacteria are grown on the surface of agar gel (an alga)

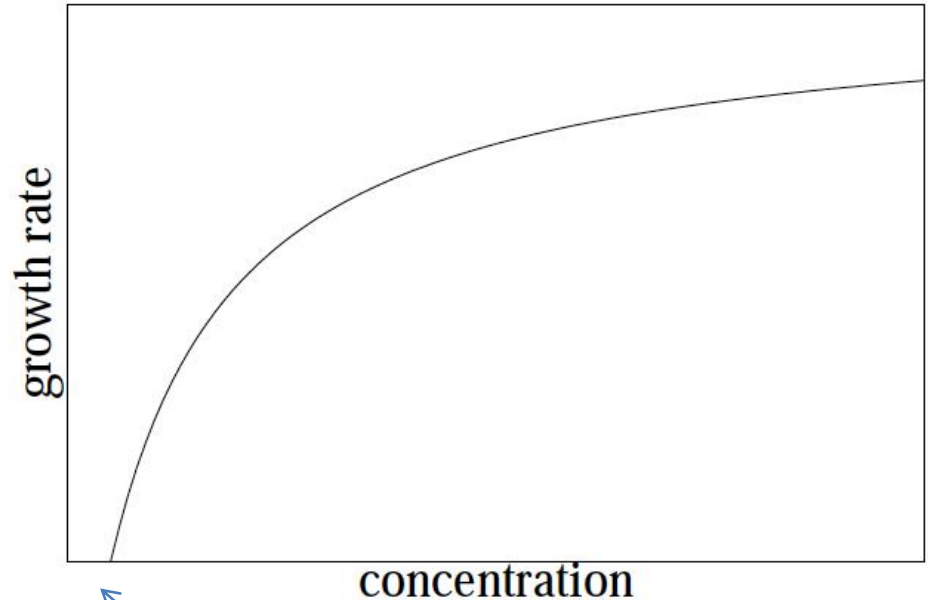
- “Dry” surface ( = big agar concentration)
  - The cells can not move (to spread over the substrate can take even weeks)
  - The duplication time is much smaller
    - *proliferation* is the key factor in determining the morphology
- “Soft” gel (= small agar conc.)

Or: the bacteria produce surfactant

- The colony spreads over the substrate in a few hours
  - *bacterial motion* and *chemotaxis* are the main factors

# Microbiological background - Proliferation

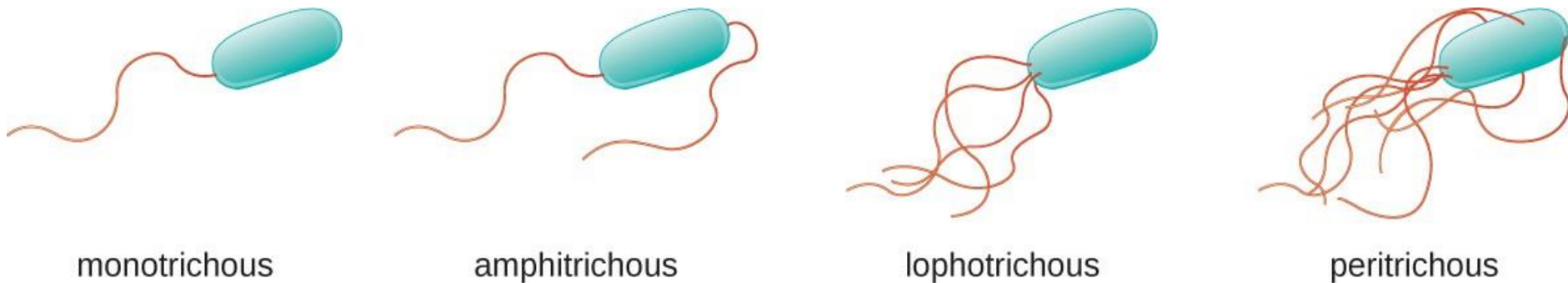
- Growth (the increase of the number and total mass of bacteria) strongly depends on the nutrient concentration
- Rate of growth (number of cell divisions within a population of unit size during a unit time interval) increases with the nutrient concentration in a hyperbolic manner.



A certain amount of nutrient is required to maintain the intracellular biochemical processes

# Microbiological background - Motility

Procaryotes move in aquatic environment by rotating their flagella



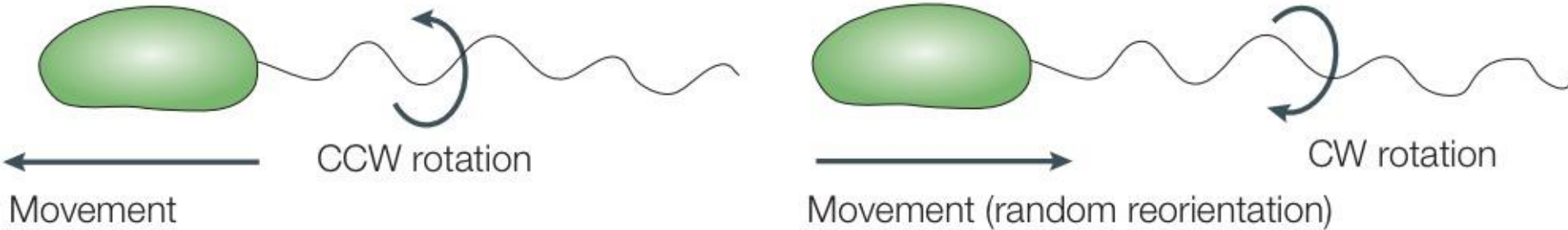
Bacteria can have

- One flagellum, “monotrichous”
- A pair of flagella at the opposite cell poles, “amphitrichous”
- Clusters of flagella at the poles, “lophotrichous”
- Uniformly distributed flagella over the cell membrane, “peritrichous”



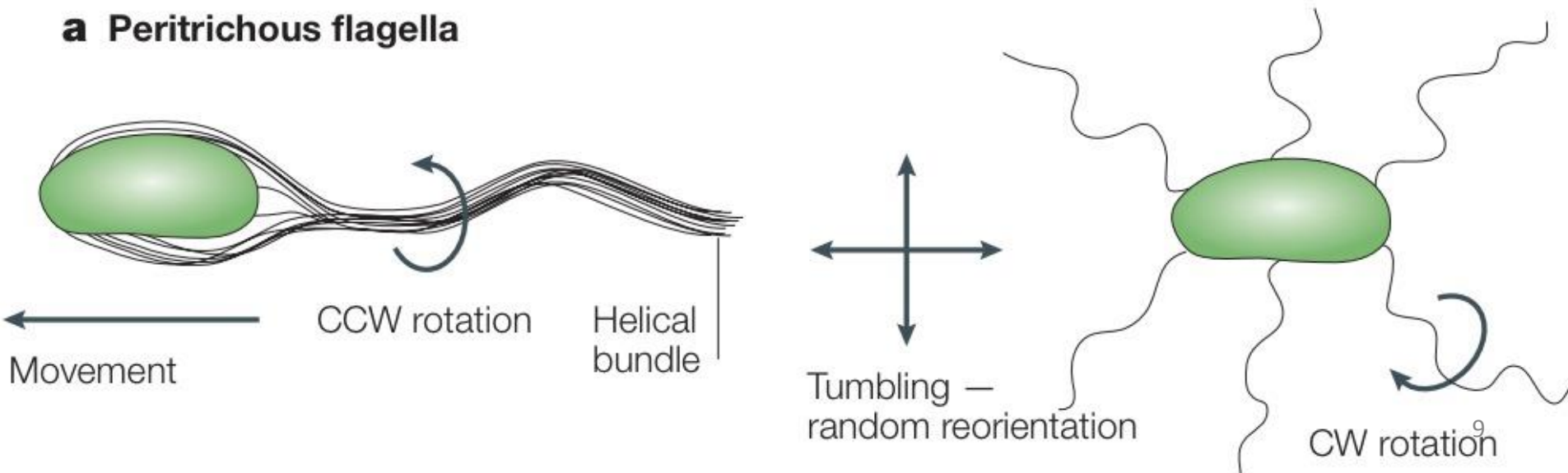
# The direction of flagellar rotation determines the motion

## **b Monotrichous flagellum**



The forward motion is interrupted by short intervals of “tumbling”

## **a Peritrichous flagella**

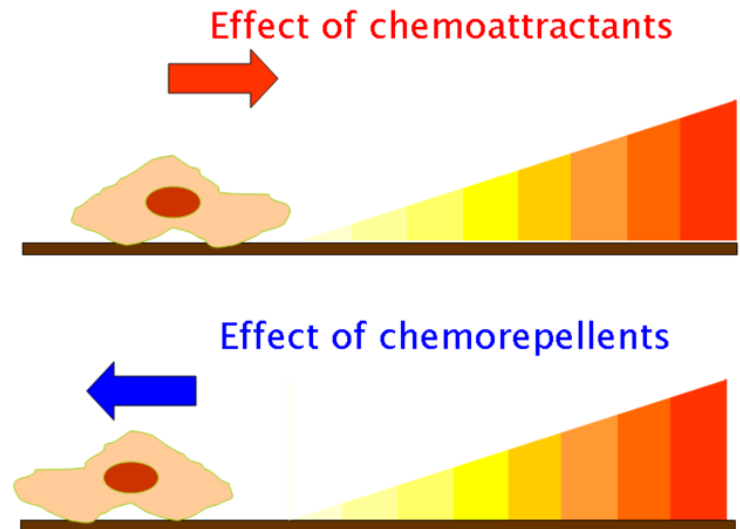


# Bacterial Motility - Gliding

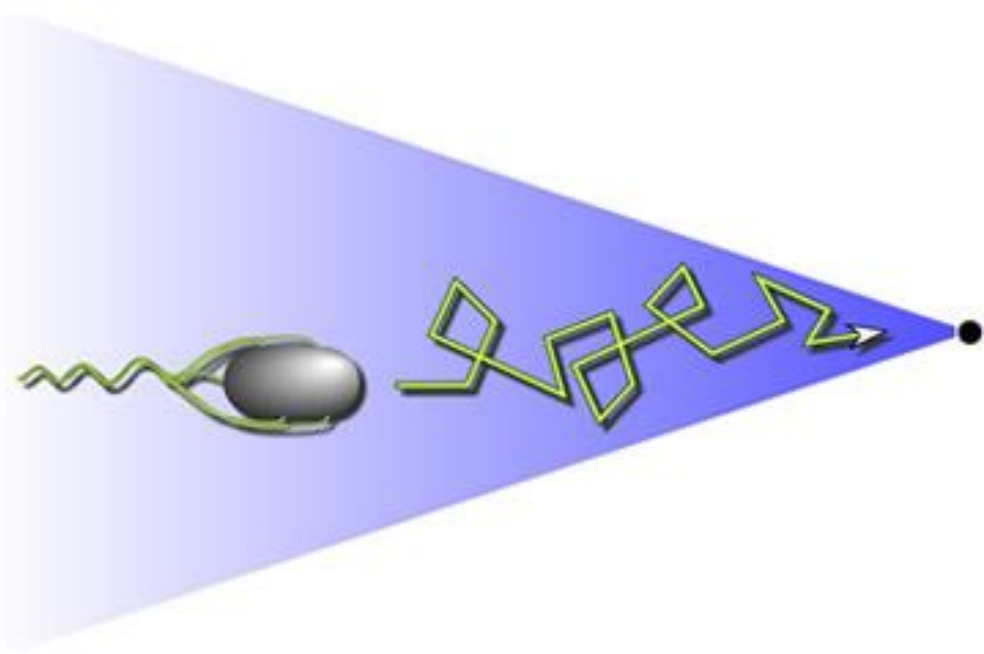
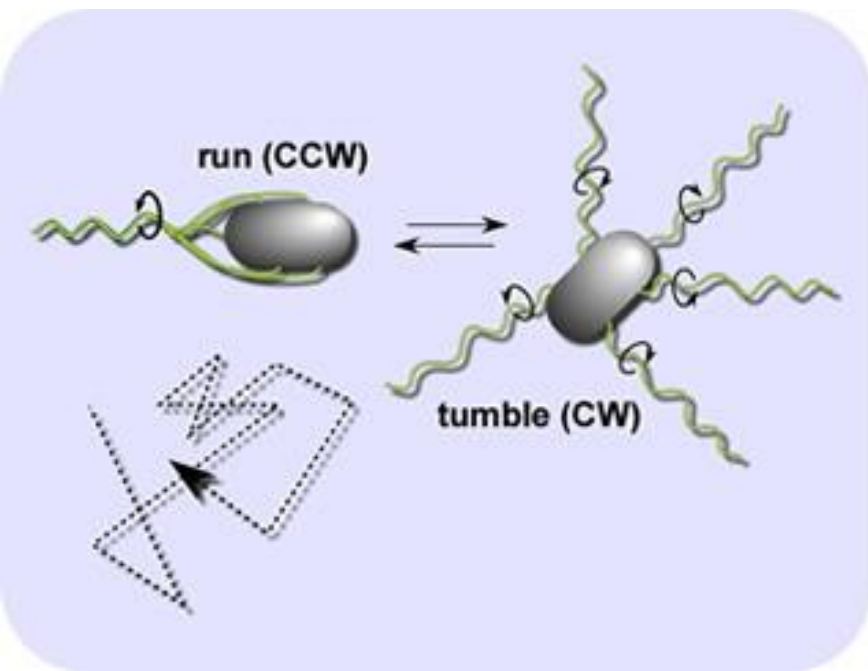
- Entirely different type of motility (flagella-independent)
- Slower and smoother than swimming
- Requires surface contact
- Employed by many strains when moving on surfaces
- No visible cellular structures associated – little is known about it
- Slime secretion
- Motion types varies greatly → probably more than one mechanisms exist
  - Gliding along the direction of the long axis of the cell (e.g. Myxococcus or Flexibacter)
  - Screw-like motion (e.g. Saprospira)
  - Direction perpendicular to the long axis (Simonsiella)

# Microbiological background - Chemotaxis

- Bacteria are attracted by nutrients (sugar, amino acids, etc.) and repelled by harmful substances and metabolic waste products.
- Other environmental factors, e.g. temperature, light, oxygen concentration
- Stochastic process: chemical gradients modulate the tumbling frequency: repressed when moving towards chemoattractants
- A molecular machinery compares the changes of the chemical concentration *in time*.



# Microbiological background - Chemotaxis



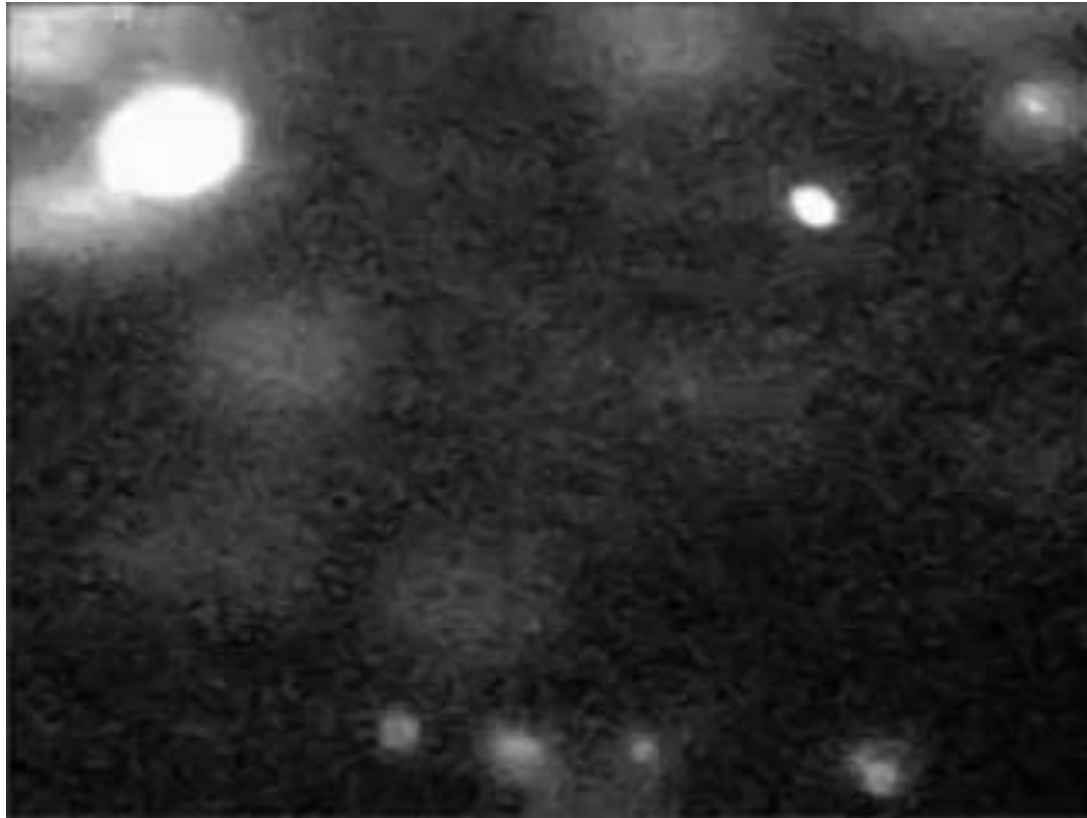
**Random and biased walks. Left:** A random walk in isotropic environments. When the cell's motors rotate CCW, the flagellar filaments form a trailing bundle that pushes the cell forward. When one or more of the flagellar motors reverses to CW rotation, that filament undergoes a shape change (owing to the torque reversal) that disrupts the bundle. Until all motors once again turn in the CCW direction, the filaments act independently to push and pull the cell in a chaotic tumbling motion. Tumbling episodes enable the cell to try new, randomly-determined swimming directions. **Right** A biased walk in a chemo-effector gradient. Sensory information suppresses tumbling whenever the cell happens to head in a favorable direction. The cells cannot head directly up-gradient because they are frequently knocked off course by Brownian motion.

# Microbiological background: Bacterial Motility. 4:35 mins



<https://www.youtube.com/watch?v=zIFRJftA2bU>

# Bacterial Flagellum. 5:30 mins



<https://www.youtube.com/watch?v=v1NnMmw8v80>

# Morphology diagram

- A diagram showing the shape (morphology ) of the bacterium colony as a function of certain environmental parameters
  - temperature, humidity, chemical composition of the substrate, etc.
  - Can result in different morphologies even for the same strain
- Characteristic colony shapes are assigned to the parameter pairs
- Most systematic experiments explore the relation between the concentration of the *agar* and *nutrients*.
- Agar concentration (consistency of the gel) determines:
  - motility of the bacteria and
  - diffusibility of the nutrient
- Nutrient concentration determines:
  - the proliferation rate

# Morphology diagram of *Bacillus Subtilis*

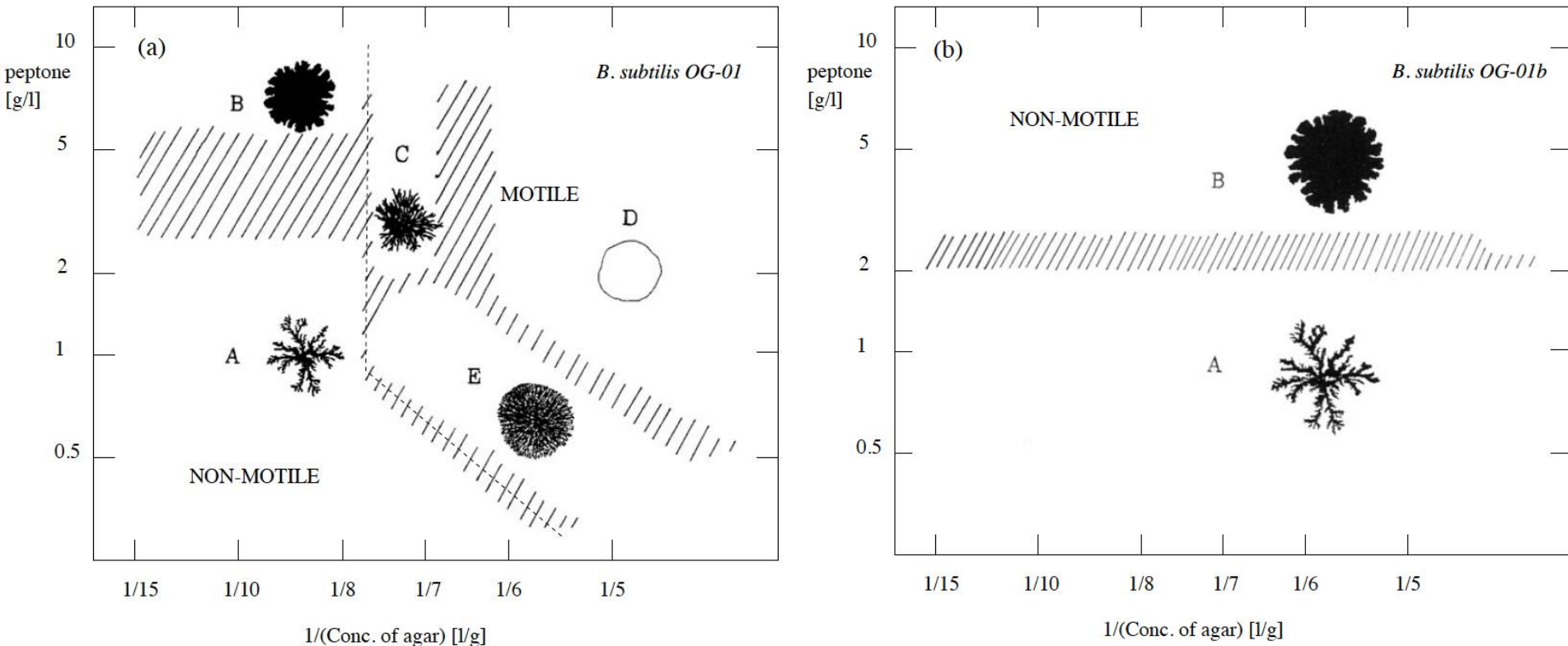
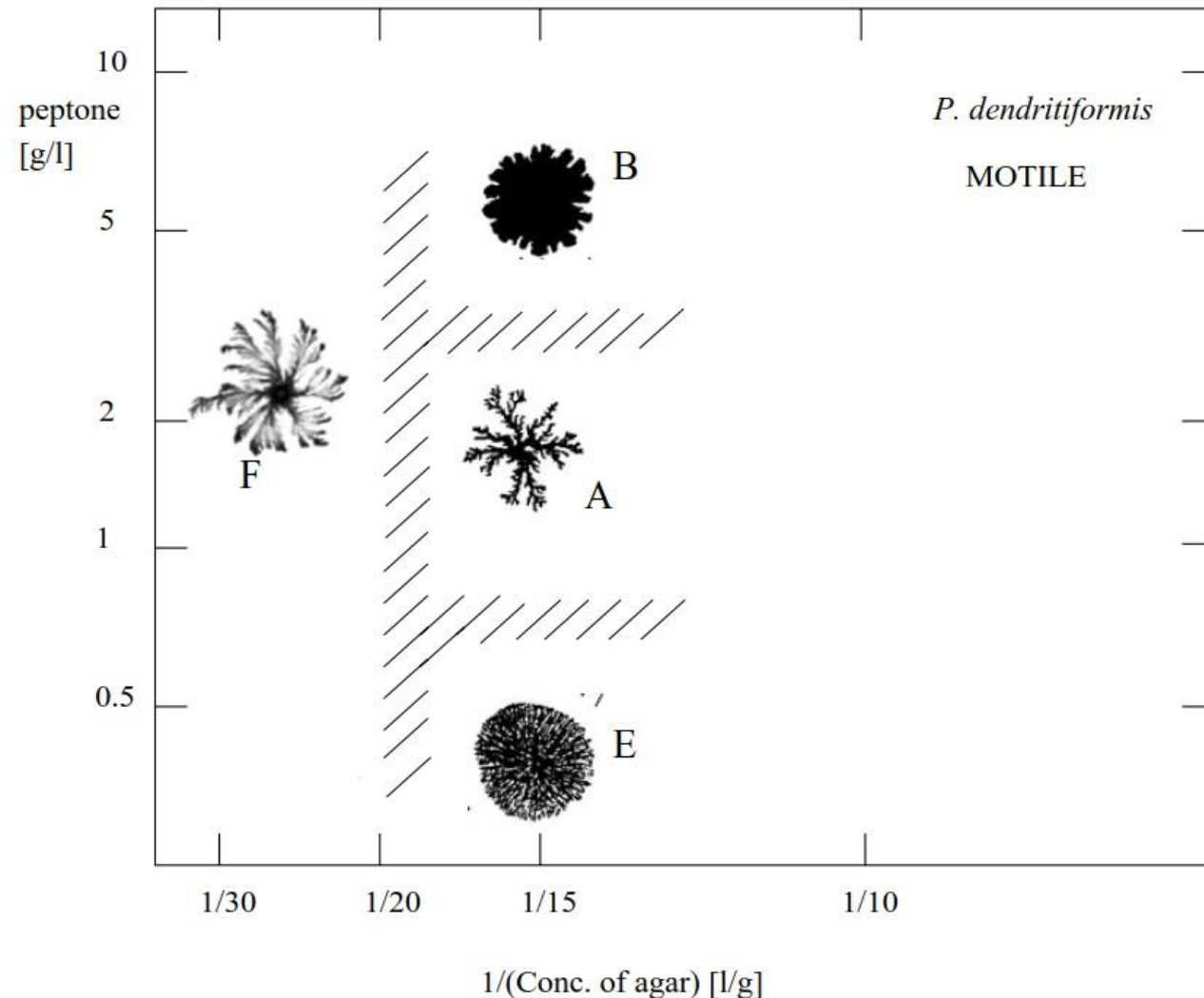


Figure 4.4: Morphology diagram of the strain *Bacillus subtilis* OG-01 (a) and a non motile mutant OG-01b (b) as a function of peptone and agar concentration. The dashed line in (a) indicates the boundary of the active movement of bacterial cells inside the colonies. The morphologies are classified as follows: fractal (A), compact with rough boundary (B), branching with periodic growth phases (C), compact with diffuse boundary (D) and dense branching (E). In the case of the non motile strain the regions A and B seen in (a) expand laterally, while regions C, D and E disappear.



# Morphology diagram of *Paenibacillus dendritiformis*



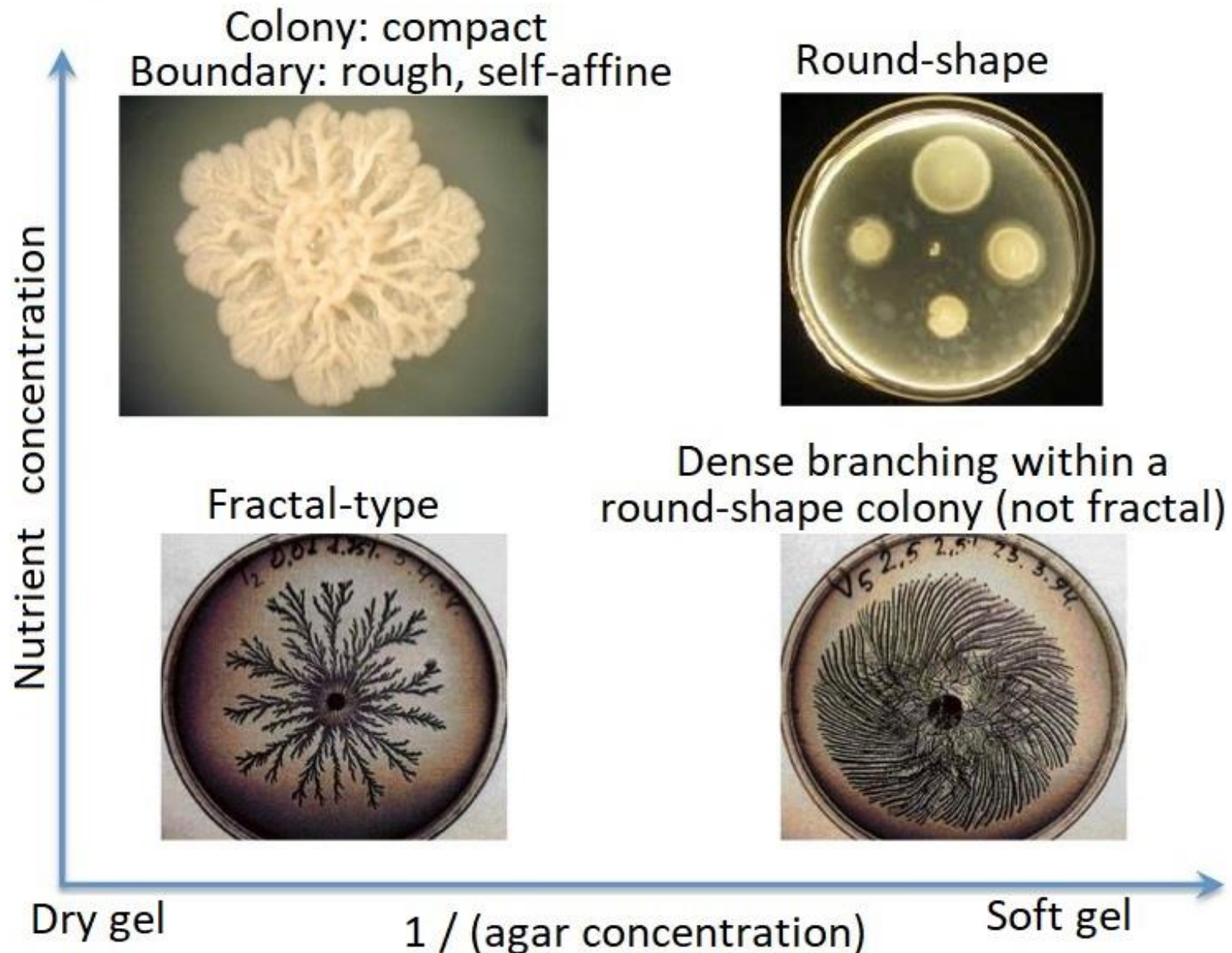
“A”: Fractal

“B”: Compact with rough boundary

“E”: Dense branching

“F”: On hard substrate a new, “twisted” morphology appears

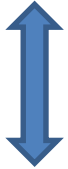
# “Summary” of the morphology diagrams



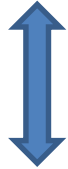
# Compact morphology

Abundant nutrient → compact colony

Either smooth or irregular perimeter



Soft gel

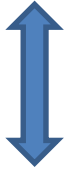


dry gel

# Compact morphology

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Either smooth or irregular perimeter



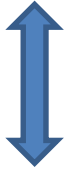
Soft gel → - Bacteria can move

- Takes a few hours to migrate across the dish
- Random walk trajectory

# Compact morphology

Abundant nutrient → compact colony

Either smooth or irregular perimeter



Soft gel → - Bacteria can move

- Takes a few hours to migrate across the dish
- Random walk trajectory


→ Inter-cellular interactions are negligible

→ Time dependence of the bacterial density  $\rho$  can be described by the Fisher-Kolmogorov equation

# Fisher-Kolmogorov equation

Starts as a small spot

- Diffuses due to random translation, and multiplies


$$\frac{\partial \rho}{\partial t} = D_{\rho} \nabla^2 \rho + f(\rho, c)$$

## Notations:

$\rho = \rho(\vec{r}, t)$  : bacterial density

$D_{\rho}$  : Diffusion coefficient (can be measured as the average displacement of the cells within a given time interval – see later)

$\nabla$  : Partial derivative with respect of the space coordinates

$f = f(\rho, c)$  : Bacterial multiplication

$c$  : Nutrient concentration

# Fisher-Kolmogorov equation – cont.

$$\frac{\partial \rho}{\partial t} = D_{\rho} \nabla^2 \rho + f(\rho, c)$$

- $D_{\rho}$ , (diffusion coefficient) can be determined from the (measurable) squared displacements  $d^2(t)$  of the individual cells during a time period  $t$  as :

$$\overline{d^2(t)} = 2D_{\rho}t$$

(overline: averaging among the cells)

- $f(\rho, c)$ : bacterial multiplication
  - When  $\rho$  is small, cells proliferate with a fixed rate  
→ exponential growth
  - In practice, even with unlimited nutrient supply, there's a certain threshold  $\rho^*$  for the density (e.g., accumulation of toxic metabolites)
  - We choose cell density units such that  $\rho^* = 1$

# Fisher-Kolmogorov equation – cont.

$$\frac{\partial \rho}{\partial t} = D_{\rho} \nabla^2 \rho + f(\rho, c)$$

- We choose  $\rho^*=1$  (threshold-density, above which cell-density can not increase)
- Growth rate decreases as  $\rho \rightarrow \rho^*=1$ , and  $f(1)=0$
- The specific form of  $f$  is unimportant, we will use

$$f(\rho, c) = R(c) * \rho(1 - \rho)$$

which satisfies the above criteria

- $R(c)$  is a function expressing  
how the proliferation depends on the  
nutrient concentration

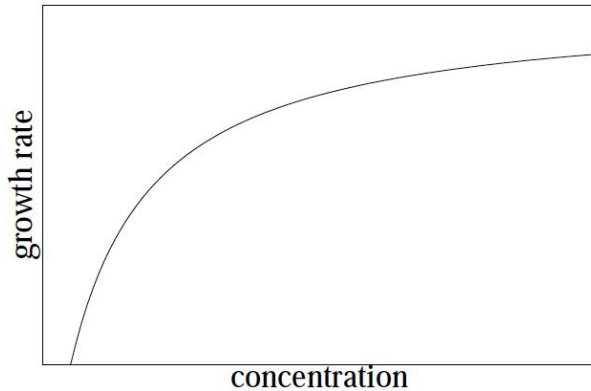
$$R(c) = \begin{cases} \text{for small } c \text{ values } R \sim c \\ \text{for big } c \text{ values } R \text{ is constant} \end{cases}$$



# Fisher-Kolmogorov equation – cont.

## Dependency on $c$ :

Hyperbolic manner



$R(c) =$  for small  $c$  values  $R \sim c$   
for big  $c$  values  $R$  is const.

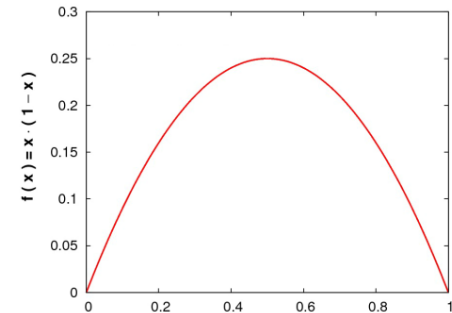
Some amount is needed for maintaining the intracellular biochemical process

## Dependency on $\rho$ :

When  $\rho$  is small, cells proliferate with a fixed rate

→ exponential growth

In practice, even with unlimited nutrient supply, there's a certain threshold  $\rho^*$  for the density (e.g., accumulation of toxic metabolites)



$$f(\rho, c) = R(c) * \rho(1 - \rho)$$

# Numerical solution of the Fisher-Kolmogorov equation in 1 D

$$\frac{\partial \rho}{\partial t} = D_\rho \nabla^2 \rho + f(\rho, c) = D_\rho \nabla^2 \rho + R(c) \rho(1 - \rho)$$

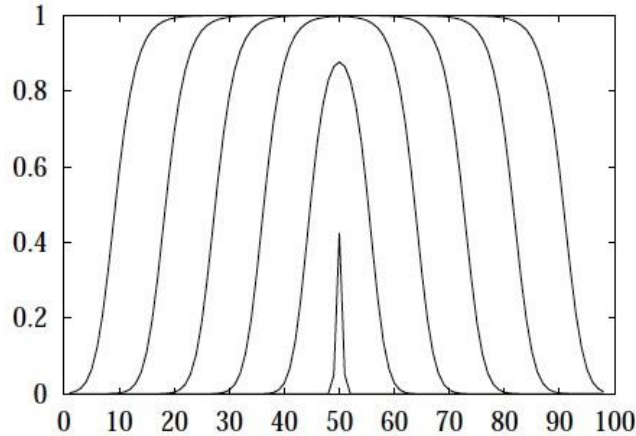


Figure 4.9: Typical result of the numerical integration of the Fisher equation (4.8) starting from a localised perturbation ( $D_\rho = 1$ ,  $f(x) = x - x^2$ , i.e.,  $r = 1$ ). The  $\rho(x, t)$  curves are plotted for  $t = 0, 5, 10, 15, 20$  and  $25$ . The domain grows with a stationary speed of  $v = 2$ .

**Numerical solution:** the growing domain of the colony expands with a constant speed  $v \approx v_*$  where

$$v_* = 2\sqrt{RD_\rho}$$

# Compact morphology

Abundant nutrient → compact colony

Either smooth or irregular perimeter



- Dry gel and/or un-motile bacteria
  - Bacteria exert mechanical pressure on their environment (in order to expand to their preferred size)
  - Inter-cellular interactions
  - Modified Fisher-Kolmogorov equation
  - Irregular (self-affine) surface

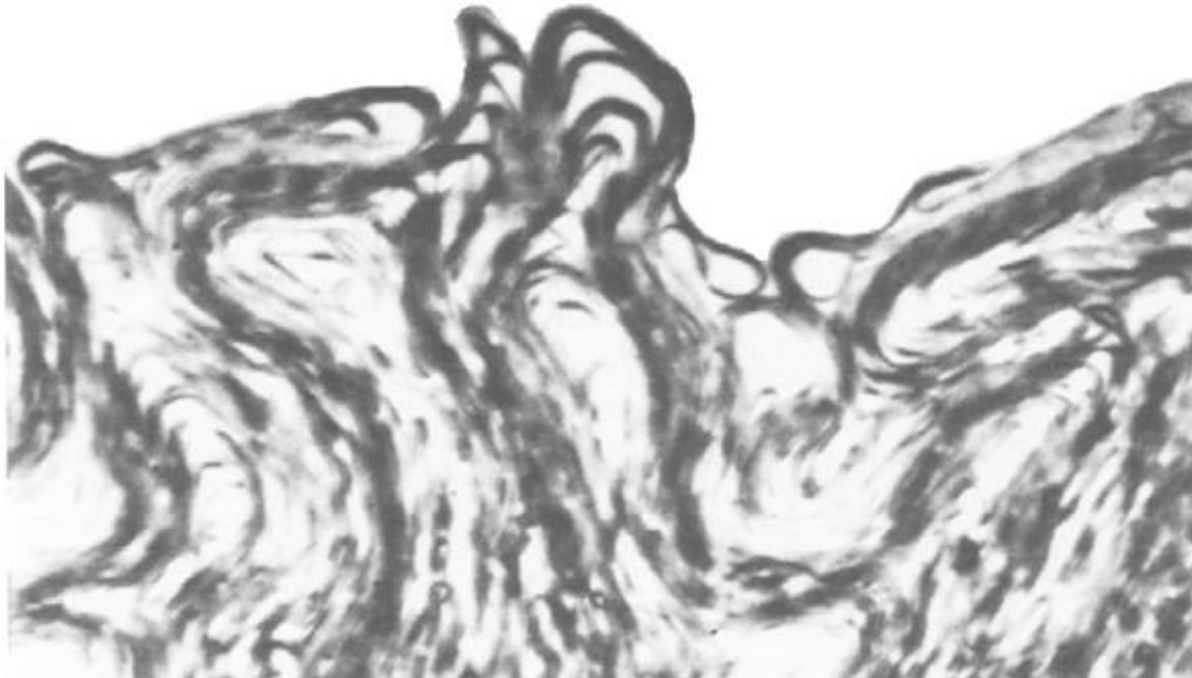
# Cell-cell interaction

- When the bacteria are *not* independent during the spreading of the colony (e.g. non-motile cells)
  - Abrupt change in the cell density at the border of the colony
- Propagation of the boundary: expansions of the cell volumes inside the colony
  - The bacteria can not expand to their preferred size, they exert mechanical pressure on their surroundings
  - Large densities:  $\rho \sim \rho - \rho_0$  ( $\rho_0$  threshold density for close-packed colonies)
- For large density values the displacement is:  $v = D_0 \nabla(\rho - \rho_0)$   
( $D_0$ : diffusion coef., similarly to  $D_\rho$  in the F-K. eq.)

- Modified F-K. eq: 
$$\partial_t \rho = \begin{cases} D_0 \nabla^2 \rho + f(\rho) & \text{for } \rho > \rho_0 \\ f(\rho) & \text{otherwise} \end{cases}$$

- In such cases the colony boundary is self -affine

# The boundary of a *Bacillus subtilis* colony (OG-01 strain) grown on hard agar



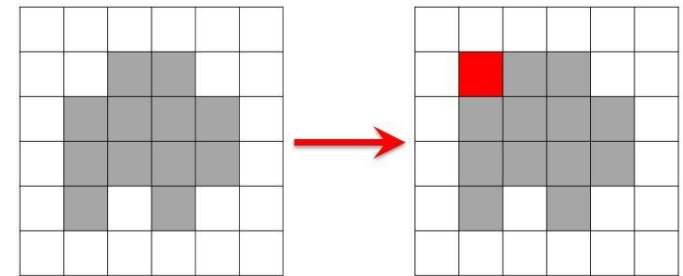
Long bundles of chains of individual cells consist the colony of the *B. subtilis* OG-01 strain grown on hard agar. Note the abrupt change in cell density at the boundary.

# The formation of self-affine boundaries – the Eden model

- One of the earliest method to generate self-affine objects (1961)
- Cells grown on a lattice
- One single rule for growing the colony:

– In each step, one of the lattice sites next to the populated areas is chosen randomly and occupied.

- Or: in each time step, a randomly chosen (non-motile) bacterium proliferates.



- Primitive, but universal model

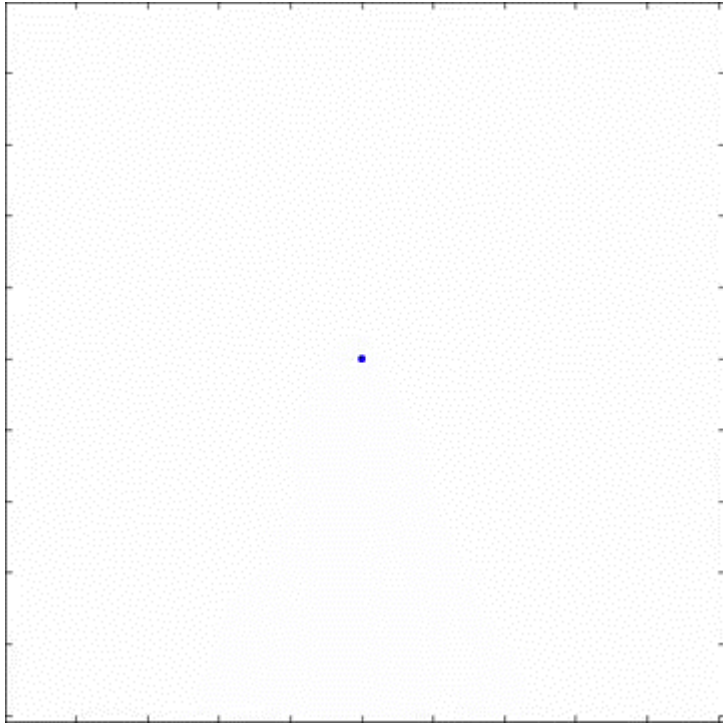
# Eden-model



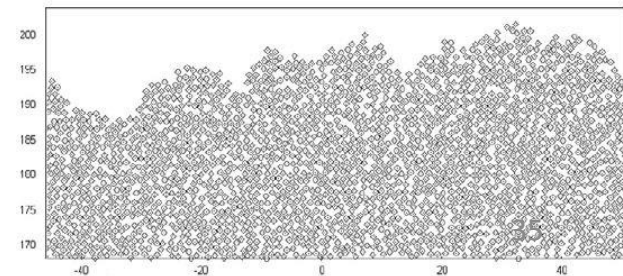
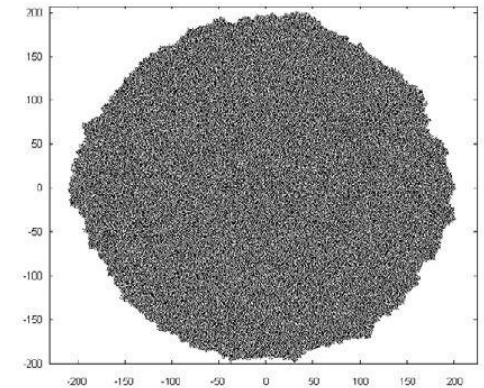
A typical colony in the Eden model grown on a strip of 256 lattice units.

- Initial step:
  - 1 occupied cell
- Variants:
  - Each position with same probability
  - Higher number of occupied neighbors increase the probability
- Variants of the model leave the statistical features of the developing clusters invariant in the asymptotic limit.

# Simulations of the Eden model in 2D



- The lattice can destroy the rotational symmetry
- Continual model is more realistic



<https://youtu.be/hluvLTwMFOs>



# Summary of the Eden model

- The surface contains “overhangs”
- Basic assumptions:
  - The units can not move (no “diffusion”)
  - Multiplication on the surface
- The model is simple but can be applied to many phenomena – “universality”
- The result is a self-affine surface
- KPZ model
  - The time evolution of the profile of a growing interface
  - Kardar, Parisi, Zhang: Dynamic scaling of growing surfaces. Physical Review Letters (1986)

# The Kardar-Parisi-Zhang (KPZ) equation

$$\partial_t h = \nu \partial_x^2 h + \frac{\lambda}{2} (\partial_x h)^2 + u + \eta$$

- $h$  : Height of the surface
- $\partial_t$  : Partial derivative with respect to time
- $\partial_x$  : Partial derivative with respect to the space coordinate  $x$ ; (  $\partial_x^2$  : second derivative )
- $\nu$  : surface tension coefficient ( $nu$ )
- $u$  : growth speed, perpendicular to the surface
- $\eta$  : uncorrelated noise (stochastic)

# The KPZ step-by-step

$$\partial_t h = \nu \partial_x^2 h + \frac{\lambda}{2} (\partial_x h)^2 + u + \eta$$

Speed of vertical growth:

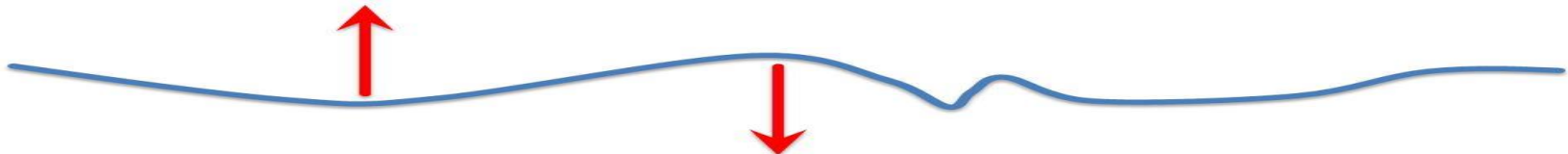
$$\partial_t h(x, t)$$

Components:

1. Surface tension term  $\nu \partial_x^2 h$

- $2^{nd}$  derivative negative  $\rightarrow$  local max (“top of a hump”)
- $2^{nd}$  derivative positive  $\rightarrow$  local min (“bottom of a swale”)
- Tends to smoothen the interface
- Does not permit discontinuities (large jumps) in  $h$
- $\nu$ : surface tension coefficient

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Speed of vertical growth:  $\partial_t h(x, t)$

2<sup>nd</sup> component: makes the surface lumpy

$$\begin{aligned} \Delta h &= \frac{u \cdot \Delta t}{\cos \varphi} = u \cdot \Delta t \left( \frac{1}{\cos \varphi} \right) = u \cdot \Delta t \sqrt{1 + \tan^2 \varphi} \approx \\ &\approx u \cdot \Delta t \left( 1 + \frac{\tan^2 \varphi}{2} \right) = u \cdot \Delta t + \frac{u \cdot \Delta t}{2} \tan^2 \varphi \approx \\ &\approx u \cdot \Delta t + \frac{u \cdot \Delta t}{2} (\partial_x h)^2 \end{aligned}$$

During a small  $\Delta t$  period of time the growth of the surface:

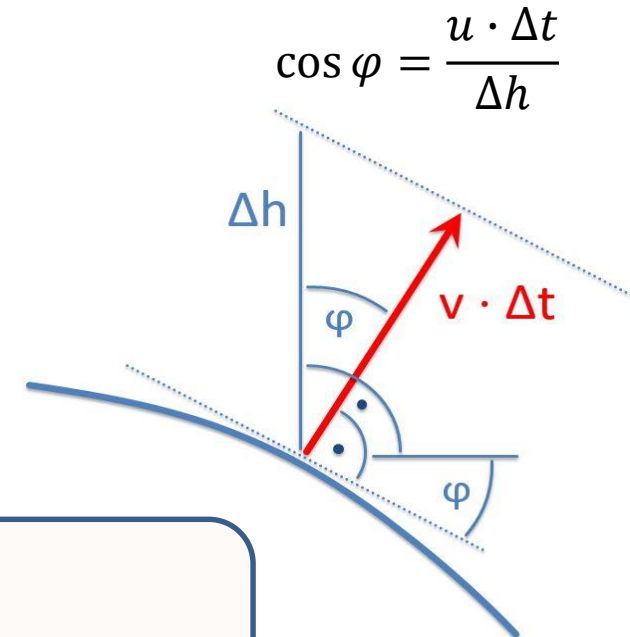
$$\frac{\Delta h}{\Delta t} \approx u + \frac{u}{2} (\partial_x h)^2$$

Due to other effects  $\frac{u}{2} \rightarrow \frac{\lambda}{2}$  (more general equation)

1D  $\rightarrow$  2D

- $x \rightarrow r$
- $\partial_x \rightarrow \vec{\nabla}$

$$\begin{aligned} 1 + \tan^2 \varphi &= \frac{1}{\cos^2 \varphi}; \\ \text{if } \varepsilon \ll 1, \text{ then } \sqrt{1 + \varepsilon} &\approx 1 + \frac{\varepsilon}{2} \\ \text{if } \varphi \ll 1, \text{ then } \tan(\varphi) &\approx \partial_x h \end{aligned}$$



# The KPZ step-by-step

## 1D→2D

$$\partial_t h = \nu \partial_x^2 h + \frac{\lambda}{2} (\partial_x h)^2 + u + \eta$$

- $h$  : Height of the surface
- $\partial_t$  : Partial derivative with respect to time
- $\partial_x$  : Partial derivative with respect to the space coordinate  $x$ ; ( $\partial_x^2$  : second derivative )
- $\nu$  : surface tension coefficient ( $\nu$ )
- $u$  : growth speed, perpendicular to the surface
- $\eta$  : uncorrelated noise (stochastic)

$$\partial_t h(\vec{r}, t) = \nu \cdot \vec{\nabla}^2 h(\vec{r}, t) + \frac{\lambda}{2} (\vec{\nabla} h)^2 + u + \eta(\vec{r}, t)$$

- Smoothing component (surface tension)
- Roughening
- noise:  $\eta = \eta(\vec{r}, t)$  : stochastic (=non-deterministic), uncorrelated in space and time

## Comments:

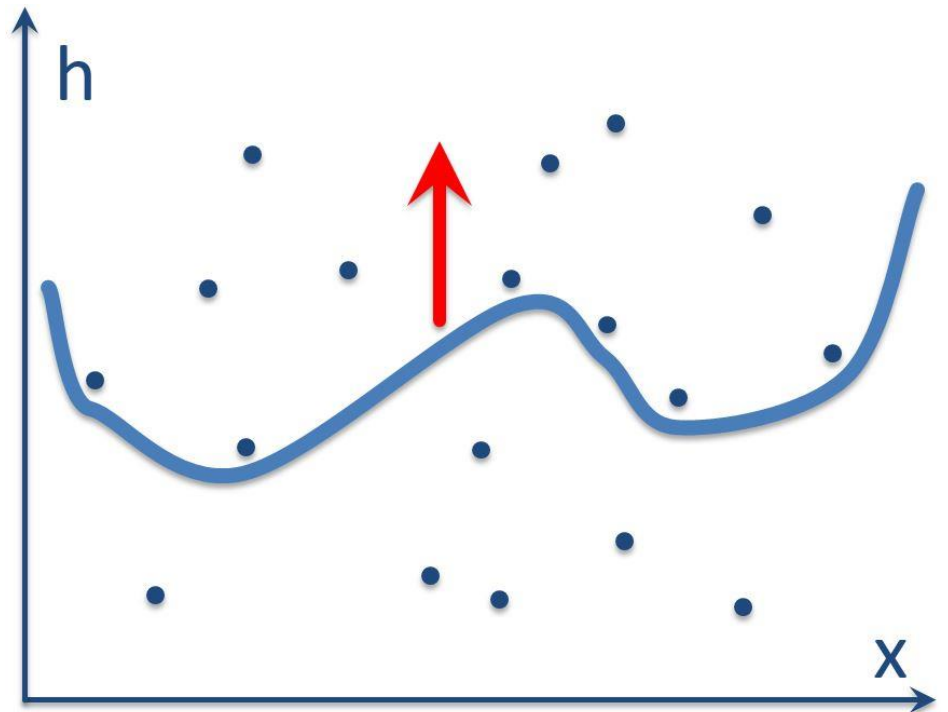
- In case of uncorrelated  $\eta(\vec{r}, t)$  noise the resulting surface is self affine
- In this case (and in the Eden model) the roughness exponent  $H=1/2$ , in contrast to experiments, where  $H \approx 0.7, \dots, 0.8$
- Reason: in the KPZ the noise is uncorrelated in time ( $\leftrightarrow$  reality!) 40

# KPZ with quenched noise

- Uncorrelated noise (in time):
  - If the noise is  $\eta(\vec{r}, t)$  at the position  $\vec{r}$  at time  $t$ , then the noise is “*independent*” of  $\eta(\vec{r}, t)$  at the same place, at time  $t + \Delta t$ .
- In other words:
  - If the spreading of the colony sticks at time  $t$  at position  $\vec{r}$  due to the local inhomogeneity  $\eta(\vec{r}, t)$  of the surface (gel), then at the same position,  $\Delta t$  later, the noise would be independent (uncorrelated), that is, the surface would move on.
- In contrast, the reality is that
  - Such noises are often constant in time
  - The colony moves in an inhomogeneous medium, in which the inhomogeneity is constant in time
  - The noise “quenches” into the medium. “quenched noise”

# KPZ with quenched noise

- If the noise is constant (and fixed) in time:
  - If the spread of the colony surface sticks at a given point  $\vec{r}$ , then this “halt” can be extensive in time, since the media does not change.
  - Results in a surface proceeding in a hopping/jiggling manner (points are blocks).



# KPZ with quenched noise

- Defining the  $\eta(\vec{r}, t)$  quenched noise:
  - Let us consider a  $\Delta(u)$  function with the following properties:
    - If  $u$  is close to 0, then  $\Delta(u) \cong 1$  (in a small, finite interval), and
    - Everywhere else  $\Delta(u)=0$ .
    - a “blurred” Dirac-delta
  - $\eta(\vec{r}, t) := 2D\tilde{\eta}(\vec{r}, h(\vec{r}, t))$ 
    - $\tilde{\eta}$  is normalized noise
    - whose spatial autocorrelation is  $C_{\tilde{\eta}}(\vec{r}, \vec{r}') = \Delta(|\vec{r}|)\Delta(|\vec{r}'|)$ 
      - That is, correlated in a very small spatial interval
    - $D$  : average magnitude of the noise as  $\sqrt{C_{\tilde{\eta}}(0,0)} = \sqrt{2D}$
  - We incorporate this quenched noise into the KPZ, we get:
$$\partial_t h(\vec{r}, t) = v \cdot \vec{\nabla}^2 h(\vec{r}, t) + \frac{\lambda}{2} (\vec{\nabla} h)^2 + u + \eta(\vec{r}, h(\vec{r}, t))$$



# KPZ with quenched noise

$$\partial_t h(\vec{r}, t) = v \cdot \vec{\nabla}^2 h(\vec{r}, t) + \frac{\lambda}{2} (\vec{\nabla} h)^2 + u + \eta(\vec{r}, h(\vec{r}, t))$$

- By “appropriate” choice of the time and length units the parameters  $\lambda$ ,  $v$  and  $u$  can be transformed out
  - the  $\lambda=v=u=1$  case:

$$\partial_t h = \vec{\nabla}^2 h + \frac{1}{2} (\vec{\nabla} h)^2 + 1 + \eta = \vec{\nabla}^2 h + \sqrt{1 + (\vec{\nabla} h)^2} + \eta$$

(where the magnitude of  $\eta$  is  $\sqrt{\eta\eta} = \sqrt{C_{\tilde{\eta}}(0,0)} = \sqrt{2D}$ )

- Two extreme cases:

1.  $D \ll D_* \sim 1$

2.  $D > D_* \sim 1$

# KPZ with small quenched noise

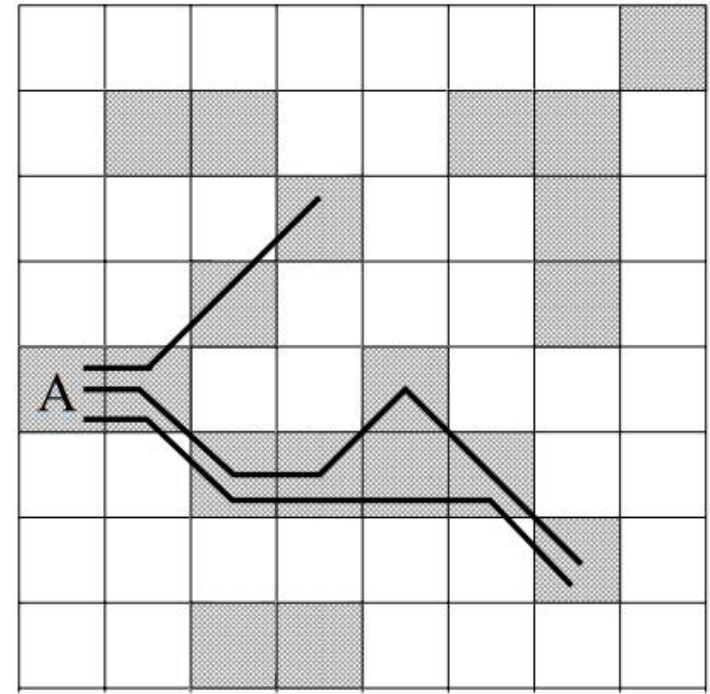
- First Case:  $D \ll D_* \sim 1$ 
  - The interface is never pinned, advances with a steady velocity
  - Fluctuating noise with some finite temporal correlations
  - The standard KPZ can be applied
  - Resulting interfaces with  $H=1/2$ .
  - Experimental support: Colonies grown on soft agar gel (small pinning effect) showed standard KPZ-like behavior with surface characterized by  $H=1/2$ .

# KPZ with big quenched noise

- Second Case:  $D > D_* \sim 1$ 
  - The interface is pinned at some certain points, for an extended period of time (until the neighboring segments pull it out)
  - If the density of the pinning points is high enough, then the propagation of the whole surface can be blocked.
  - The shape of the frozen colony is determined by the distribution of these pinning sites (and independent of the growth dynamics).
- The surface roughening can be mapped onto a **directed percolation problem**:
  - finding directed and connected paths
  - Let us consider a **lattice** instead of the continuous case (discrete model, regarding both  $h$  and the location  $(x,y)$  )

# Directed percolation

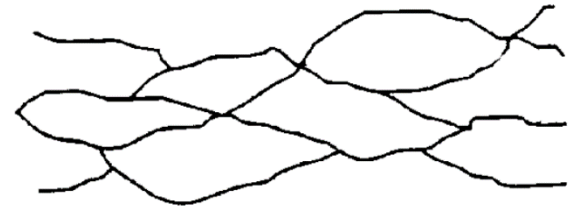
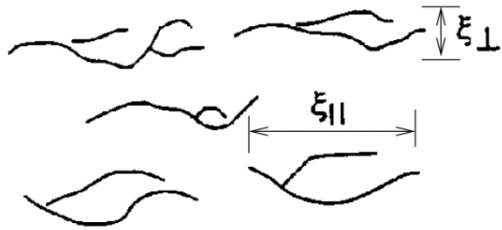
1. Let us define each lattice site as
2. “pinning” with a probability  $0 < p < 1$ . (gray squares)
3. We start from one end of the panel
4. On the pinning sites we can move ahead, up and down (but not backwards)
5. Do we reach the other end of the board?



- The chain of the pinning sites define a directed percolation cluster (if it exists).
  - Complete blocking of the interface propagation appears when there is a directed, connected path (a directed percolation cluster)
  - The propagation stops along these clusters

# Correlation lengths of directed percolation clusters

- DPC is characterized by two correlation length:
  1. Parallel to the interface (to the preferred direction)  $\xi_{\parallel}$
  2. Perpendicular to the interface (to the preferred direction)  $\xi_{\perp}$



- There is a critical probability  $p_c$  (defining the density of the pinning sites)

$$\xi_{\parallel} \sim |p - p_c|^{-\nu_{\parallel}} \quad \text{and} \quad \xi_{\perp} \sim |p - p_c|^{-\nu_{\perp}}$$

with

$$\nu_{\parallel} = 1.733 \text{ and } \nu_{\perp} = 1.097 \quad (\text{numerical results})$$

- The width of the interface:  $w \cong \xi_{\perp}$
- Complete blocking of the interface when  $\xi_{\parallel} = L$

( $L$  is the system size)

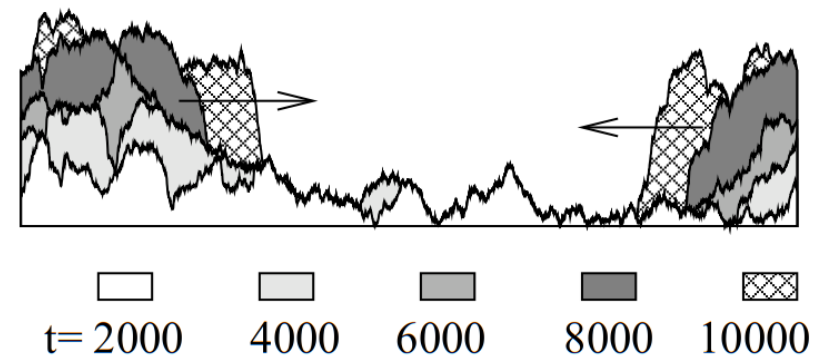
- The width of the interface at the critical point:

$$L^H \sim w \cong \xi_{\perp} \sim |p - p_c|^{-\nu_{\perp}} \sim \xi_{\parallel}^{\frac{\nu_{\perp}}{\nu_{\parallel}}} \approx L^{\frac{\nu_{\perp}}{\nu_{\parallel}}} \quad \rightarrow$$

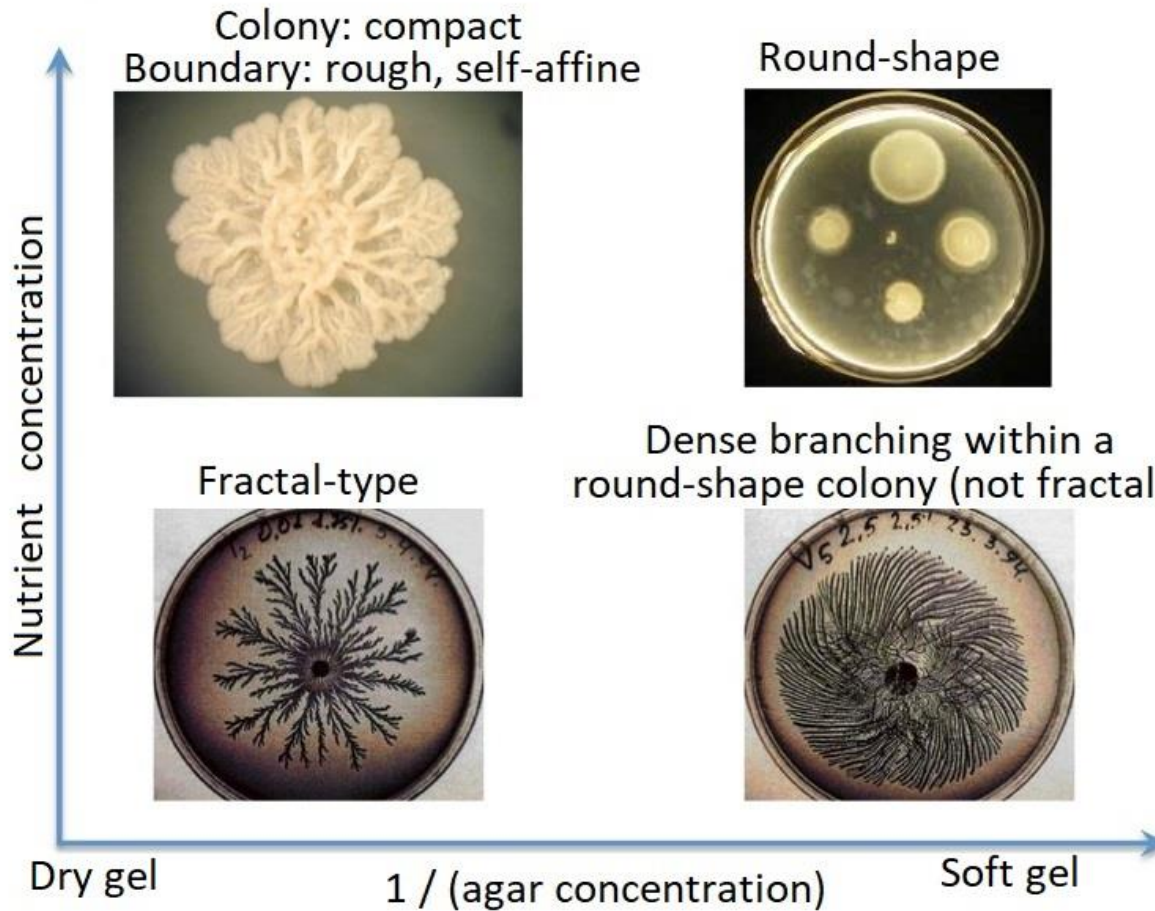
$$H = \frac{\nu_{\perp}}{\nu_{\parallel}} = 0.633$$

# Directed percolation

- The numerical result for the (discrete) directed percolation problem is  $H=0.633$ 
  - (complete blocking)
- Experimental results:  $H \approx 0.7-0.8$
- Reason: the observed colonies have both blocked and freely moving parts  $\rightarrow$  *higher* roughness exponent ( $H$ ) than for the blocked interface.
  - Numerical simulations:  
 $H \approx 0.71-0.75$   
(close to the observations)
- KPZ with quenched noise and the DP simulations have the same results



# Branching morphology



*Baillus subtilis* colony, under nutrient-poor conditions. 8 and 19 days after inoculation.

- Nutrient-poor agar substrate
- Complex, branching colonies
- Not exhibited by all strains (but by many)

# Branching morphology – colony formation

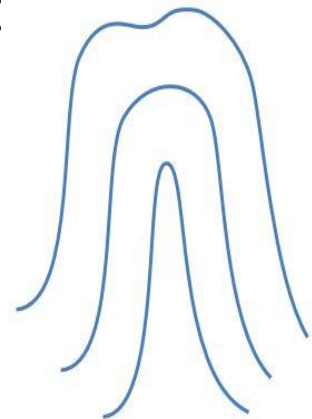
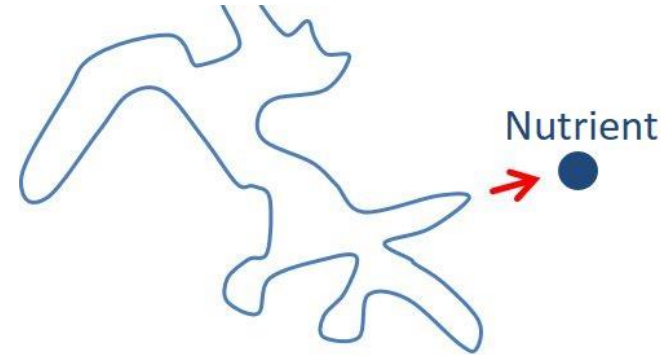
- Basic assumption:
  - the growth of the colony is *diffusion-limited*:
  - The multiplication of the bacteria is determined by the locally available nutrient
    - At the beginning: local nutrient is enough to maintain the growth
    - After some bacterial multiplication, nutrient deprivation progresses in and around the colony
    - Further growth depends on the diffusive transport from distant regions of the petri-dish
  - Experimental support
    - Non motile *B. subtilis* grows only towards nutrient-rich regions





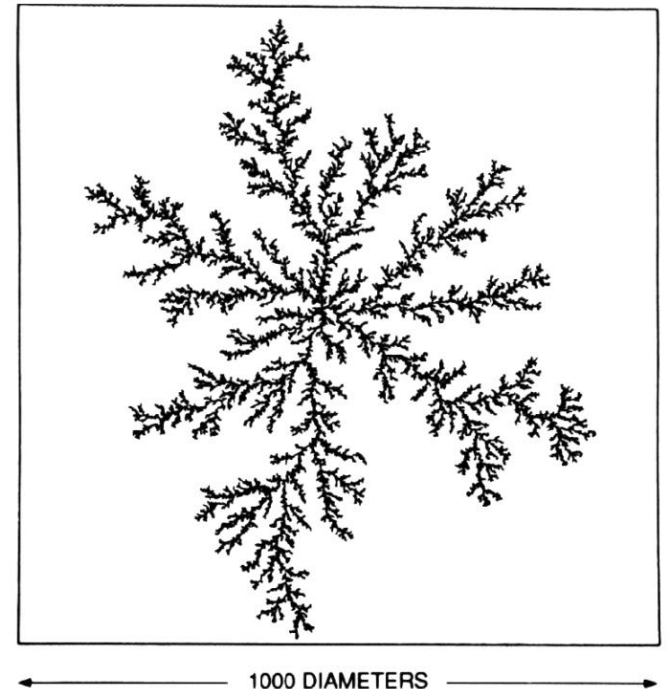
# Branching morphology – colony formation

- The speed of the growth is determined by the nutrient diffusion
- The colony develops towards the nutrient
- Instability:
  - Due to some random perturbation a small part of the colony advances “ahead” (towards some nutrient)
  - This part of the colony gets closer to the nutrient
  - Can multiply faster
- This process stops at a certain curvature
  - Certain amount of neighboring cells are needed
  - A certain “steady shape” is set
- New perturbation: new branch

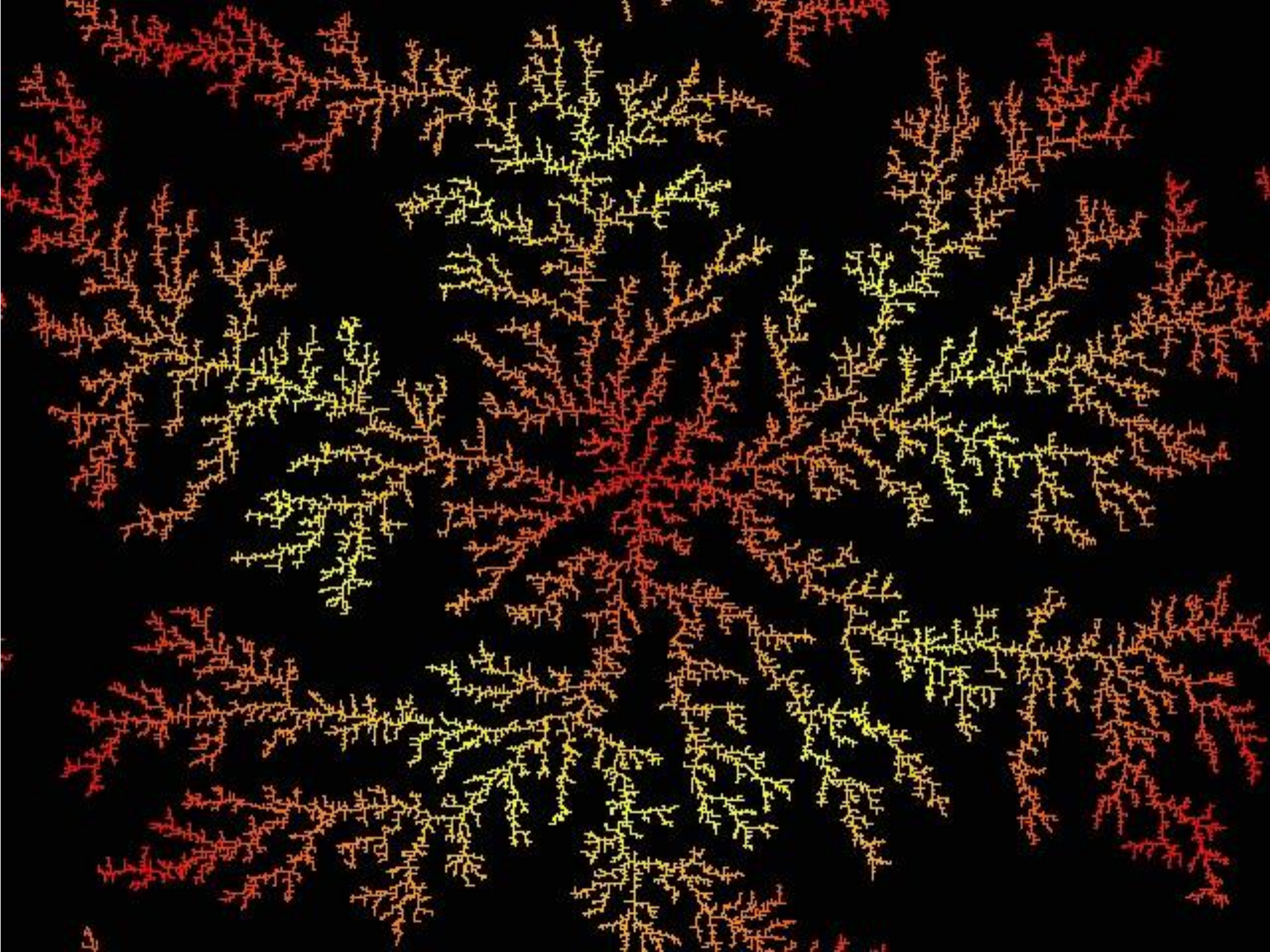


# Diffusion-Limited Aggregation (DLA)

- The definition of the basic DLA algorithm:
  - Start: 1 cell
  - In each time step:
    - A particle (performing random walk) departs from infinity (in simulations from finite distance)
    - Sticks to the colony upon graze
- Result: Fractal-type clusters

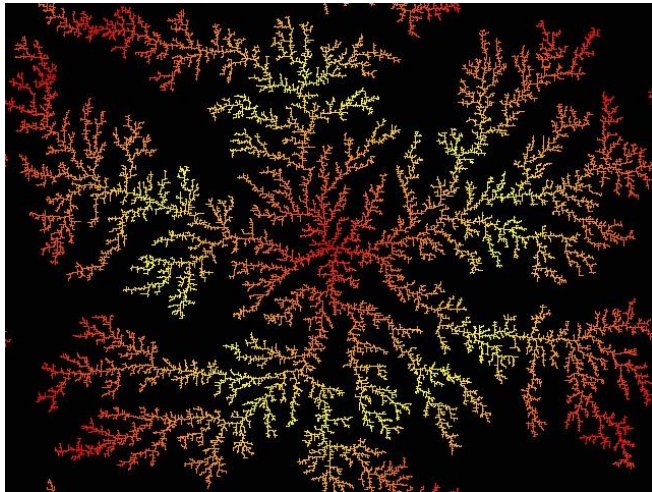


Typical DLA cluster with 50, 000 particles





# Relation to bacterium colonies




- Random walk of the particle  $\sim$  diffusion of the nutrient
- Sticking to the colony  $\sim$  bacterium proliferation
- Non-motile bacteria!
- Very simple model (1 “nutrient-unit” = 1 multiplication) generating realistic formations  $\rightarrow$  “universality”

# Refinement of the DLA model – Modeling non-motile bacteria

- Assumptions:
  - Bacteria interact with each other
  - Each particle (cell) is characterized by
    - Space coordinate  $x_i$
    - Energy state  $E_i$  (or cell cycle state)
      - $E_i < 0$  : spore state. Without nutrient, remains in this state
      - $0 < E_i < 1$  : right after multiplication
      - $E_i > 1$  : has enough energy to multiply
- Notations:
  - $\omega_i$  : nutrient consumption rate
  - $\kappa$  : conversion factor relating the maximal nutrient consumption rate with the shortest cell cycle time  
(nutrient  $\rightarrow$  energy conversion)
  - $\epsilon$  : generic “maintenance” term (not directly contributing to growth)

The energy-level of bacterium  $i$ : uptake - consumption

$$\frac{dE_i}{dt} = \kappa \cdot \omega_i - \epsilon$$


# Modeling non-motile bacteria – limits of the nutrient uptake

- Further notations:
  - $\omega_{max}$  : maximal nutrient uptake rate of the cells
  - $\kappa$  : efficiency of the enzymatic reaction converting the nutrient into energy
  - $c(x_i)$  : nutrient concentration (around cell  $i$ )
  - $\rho(x_i)$  : local cell density
  - $\omega_0 c$  : maximal diffusive transport from the substrate to the cell
  - $\omega_i$  : nutrient consumption rate (of bacterium  $i$ )
- The rate with which the cell-mass grows:


$$\rho(x_i)\omega_i = \min[\omega_{max}\rho(x_i), \omega_0 c(x_i)]$$



→ The nutrient-uptake is limited by the enzymatic rates and local nutrient concentration (the maximal speed with which cell  $i$  can take in the nutrient)

# Modeling non-motile bacteria – How the local nutrient concentration varies

- Bacteria use up the nutrient
- Changes in  $c$  are given by the diffusion equation with the appropriate sink terms at the position of the active particles:

$$\frac{dc}{dt} = D_c \nabla^2 c - \sum_i \omega_i \delta(x - x_i)$$

  
Diffusion

  
Sinks: the cell at location  $x_i$  consumes the nutrient with rate  $\omega_i$

# Summary: non-motile bacteria in nutrient-poor environment

(i) The energy-level of cell  $i$

$$\frac{dE_i}{dt} = \kappa \cdot \omega_i - \epsilon$$

- $E_i$  : energy level of cell  $i$
- $\kappa$  : efficiency of the enzymatic reaction converting the nutrient into energy
- $\omega_i$  : nutrient consumption rate
- $\epsilon$  : generic “maintenance” term (not directly contributing to growth)
- $\rho(x_i)$  : local cell density
- $\omega_{max}$  : maximal nutrient uptake rate of the cells
- $c(x_i)$  : nutrient concentration (around cell  $i$ )
- $\omega_0 c$  : maximal diffusive transport from the substrate to the cell
- $D_c$  : nutrient diffusivity

(ii) Cell-mass growth rate

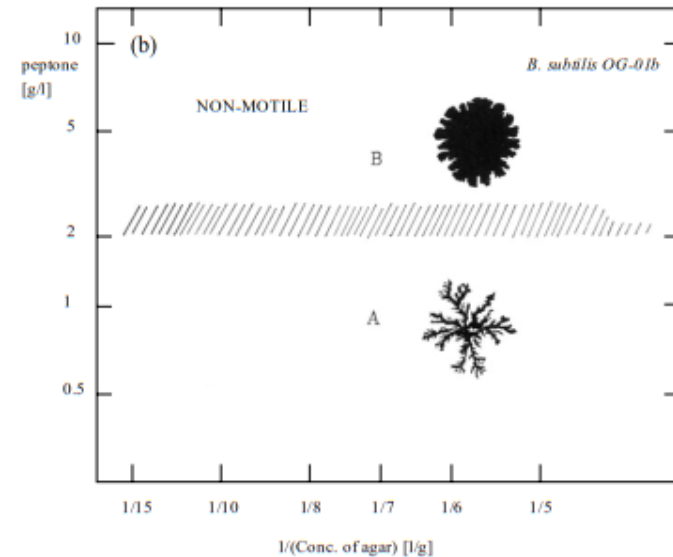
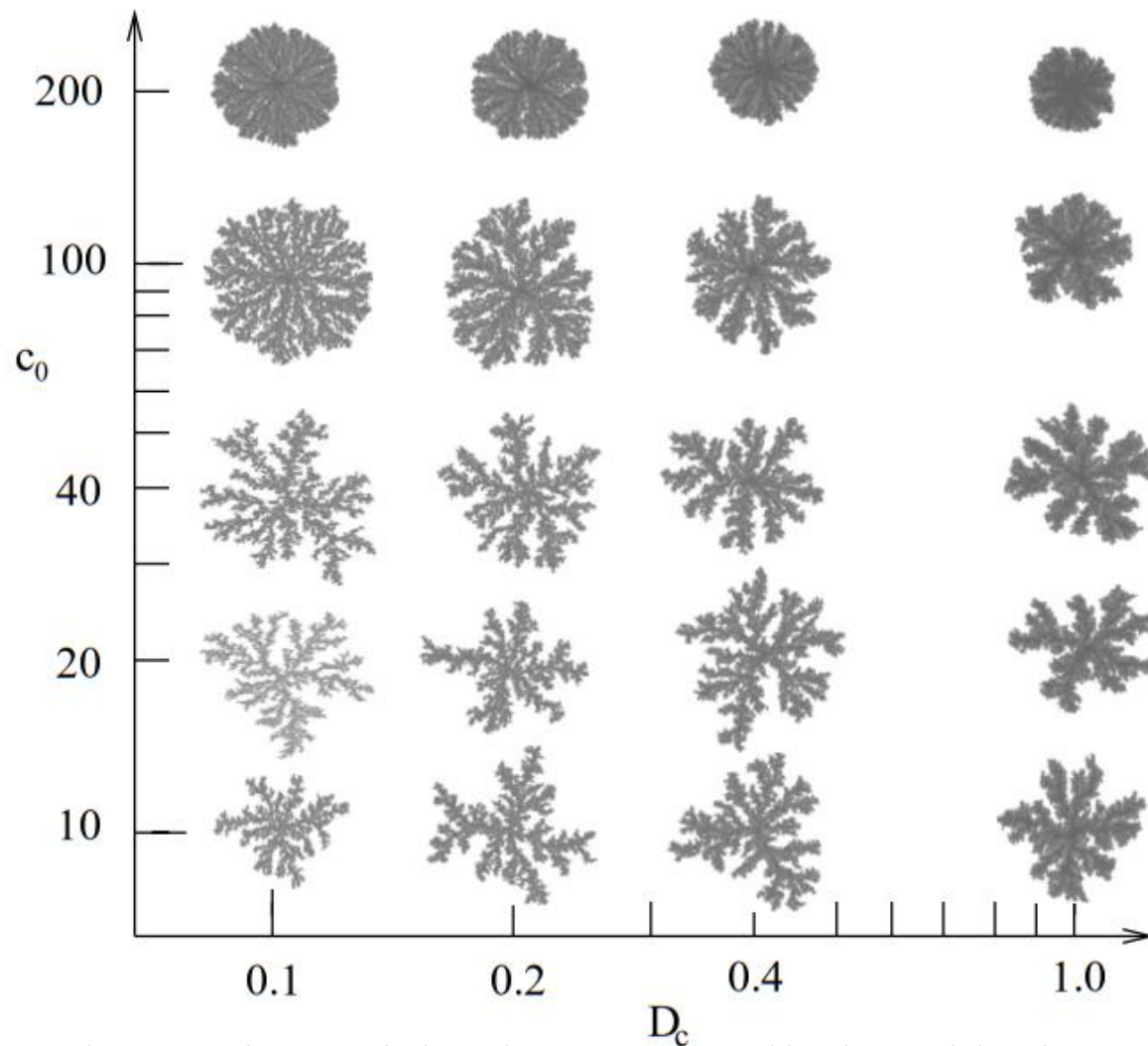
$$\rho(x_i)\omega_i = \min[\omega_{max}\rho(x_i), \omega_0 c(x_i)]$$

(iii) Changes of the local nutrient concentration

$$\frac{dc}{dt} = D_c \nabla^2 c - \sum_i \omega_i \delta(x - x_i)$$



# Results: Modeling non-motile bacteria with the refined DLA model



Related experiments: the morphology diagram of the non-motile *B. subtilis* OG-01b strain

Simulation results: Morphology diagram generated by the model with non-motile particles as a function of the initial nutrient concentration ( $c_0$ ) and nutrient diffusivity ( $D_c$ ). The colonies were grown (in the computer) until either their size or the number of bacteria reached a threshold value.

# Motile bacteria with the DLA model

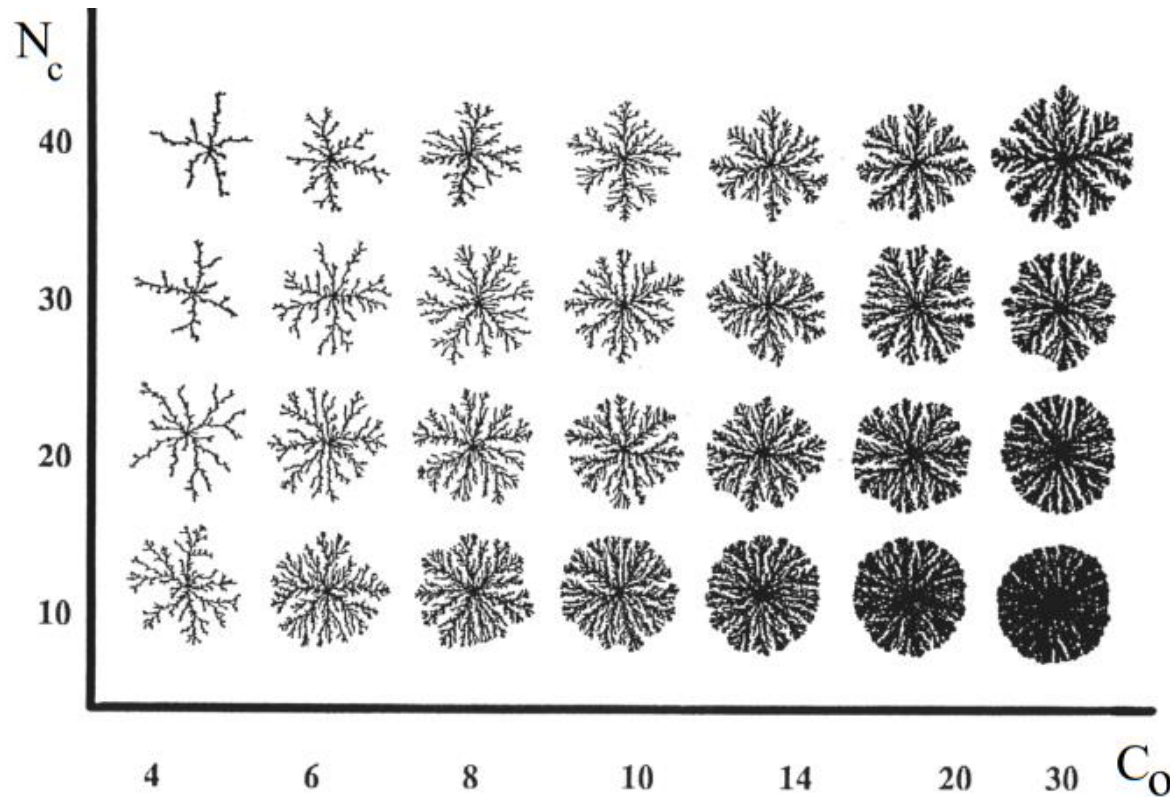
- Rules (i)-(iii) remain the same
  - (i) energy-level of cell  $i$
  - (ii) cell-mass growth rate
  - (iii) changes of the local nutrient concentration
- Non-motile cells  $\rightarrow$  self-affine surface
- Motile bacteria  $\rightarrow$  smooth surface. But they still can not migrate on dry agar surface, only where some surfactant was secreted:
- New rules:
  - (iv) The active particles move randomly (with Brownian motion) within a boundary:

$$\frac{dx_i}{dt} = v_o \vec{e}$$

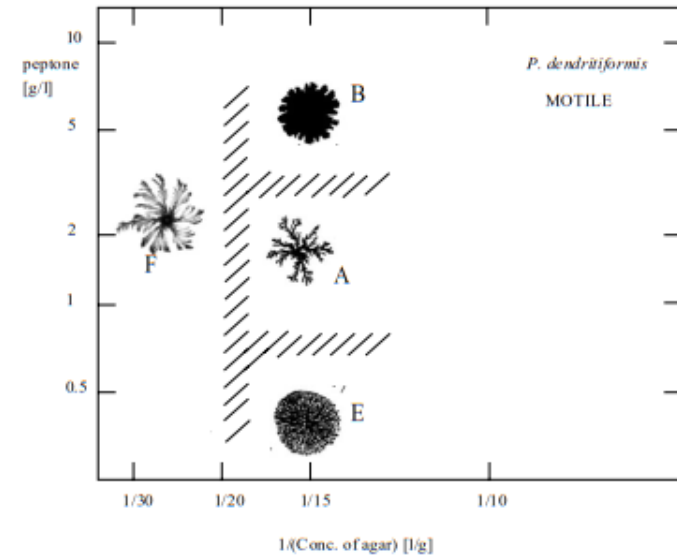
where  $\vec{e}$  is a unit vector pointing in a random direction

- (v) The propagation of the bacteria is assumed to be proportional to the local density of the active cells. Collisions of the particles with the boundary is counted, and when a threshold value ( $N_c$ ) is reached, the neighboring cell is occupied as well. (the boundary shifts forward)

# Results: Modeling motile bacteria with the refined DLA model (on hard agar gel)



Morphology diagram generated by the model with motile bacteria as the function of the initial nutrient concentration ( $C_0$ ) and agar gel “hardness”, (the threshold value  $N_c$  for the colony borderline displacement).



Corresponding experimental results: Morphology diagram of *Paenibacillus dendritiformis*. Agreement with the model results within a limited region of the parameters, but it fails to predict the formation of the thin, straight radial branches at very low food concentrations. 62

# Simulation results:

- Nice agreement with the experiments, but
- Fails to explain the transition between the  
Fractal-type and non-fractal-type colonies



Solution: assuming repulsive chemotaxis signaling among the cells. Due to the repulsion the cells by-pass each other: the random Brownian motion becomes biased.

Simulation results without  
(a) and with (b) repulsive  
chemotaxis signaling.

