

Simulations in Statistical Physics

Course for MSc physics students

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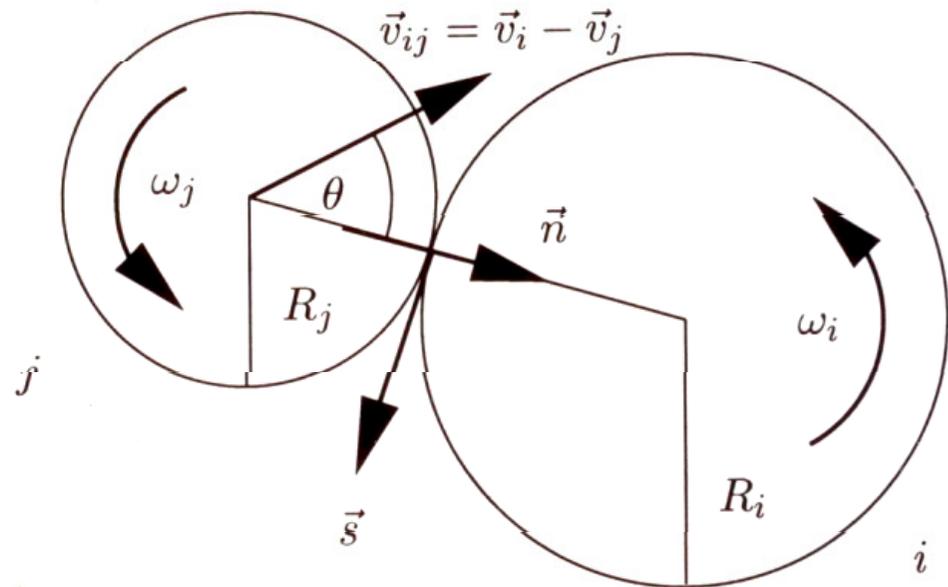
Lecture 11

Simulation of granular material (molecular dynamics = granular dynamics = **distinct element method**)

- Macroscopic particles (have temperature, pressure)
- Short range asymmetric interaction
- Dissipative collisions
- Friction

Rigid particle model: Angular momentum and friction has to be considered

Event driven



Perfect slip (no role of angular momentum)

Define „effective mass“:

$$m_{eff} \equiv \frac{m_i m_j}{m_i + m_j}$$

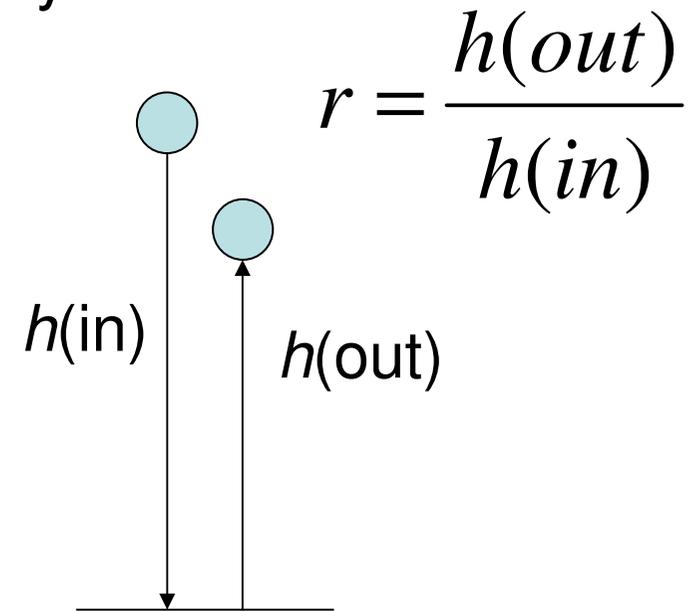
$$\overrightarrow{\Delta p}_n = -2m_{eff} \left[\left(\vec{v}_i^{before} - \vec{v}_j^{before} \right) \vec{n} \right] \vec{n}$$

$$\vec{v}_i^{after} = \vec{v}_i^{before} - \frac{\overrightarrow{\Delta p}_n}{m_i}, \quad \vec{v}_j^{after} = \vec{v}_j^{before} + \frac{\overrightarrow{\Delta p}_n}{m_j}$$

For stick condition conservation of angular momentum has to be considered

Inelastic collisions are characterized by the **restitution coefficient**

$$r = \frac{E(out)}{E(in)} = \left(\frac{v(out)}{v(in)} \right)^2 = e_n^2$$

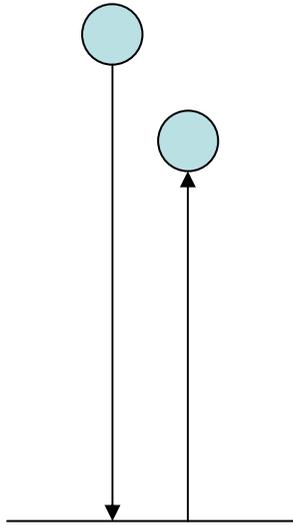


Similarly, a coefficient for the tangential component can be introduced.

$$(\mathbf{v}_j^{after} - \mathbf{v}_i^{after}) \cdot \mathbf{n} = -e_n (\mathbf{v}_j^{before} - \mathbf{v}_i^{before}) \cdot \mathbf{n}$$

$$\Delta \mathbf{p} = -m_{eff} (e_n + 1) [(\mathbf{v}_j^{before} - \mathbf{v}_i^{before}) \cdot \mathbf{n}] \cdot \mathbf{n} \quad \text{for perfect slip}$$

Finite time singularity (inelastic collapse: numerical instability):



How long does it take until it stops?

$$t_{tot} = \sum_j t_j = 2\sqrt{\frac{2h^{initial}}{g}} \sum_j \sqrt{r^j} = 2\sqrt{\frac{2h^{initial}}{g}} \left(\frac{2}{1-\sqrt{r}} - 2 \right)$$

∞ number of collisions in finite real time.
Event driven simulation does not work.

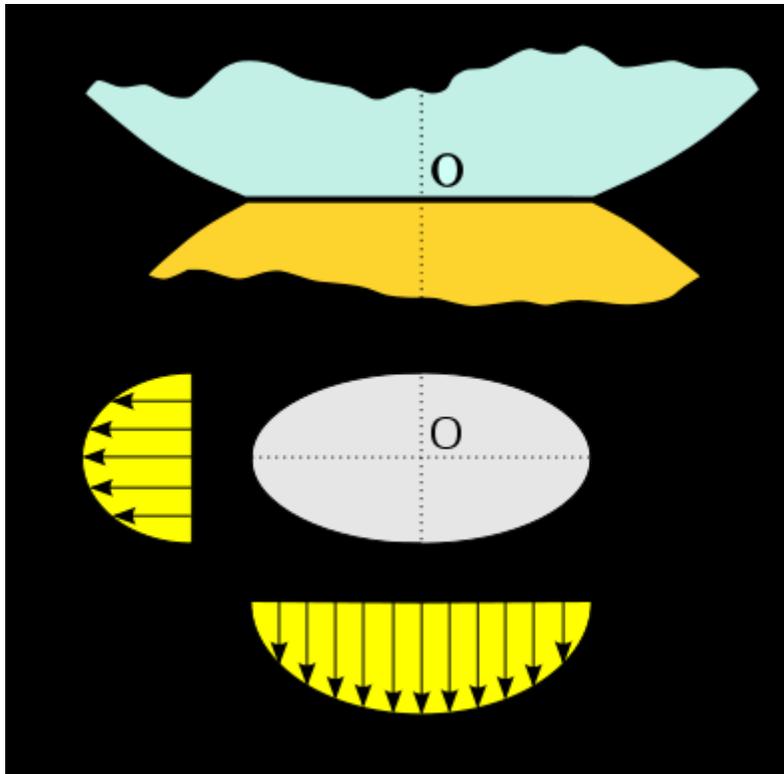
http://www.haverford.edu/physics-astro/Gollub/vib_granular/inelastic/inelastic.html

Ways out:

- Change r with the velocity (physics) $r \sim (\Delta v)^{1/5}$
- Use soft potential and time steps
- Use different technique (Contact Dynamics)

Soft potential with finite time steps

Viscoelastic model: $F = \theta(-x)(-kx - \gamma\dot{x})$



Hertzian contact: Even material with linear properties will be nonlinear. For pure elasticity:

$$F_n^{elastic} = A \xi^{3/2} \theta(\xi)$$

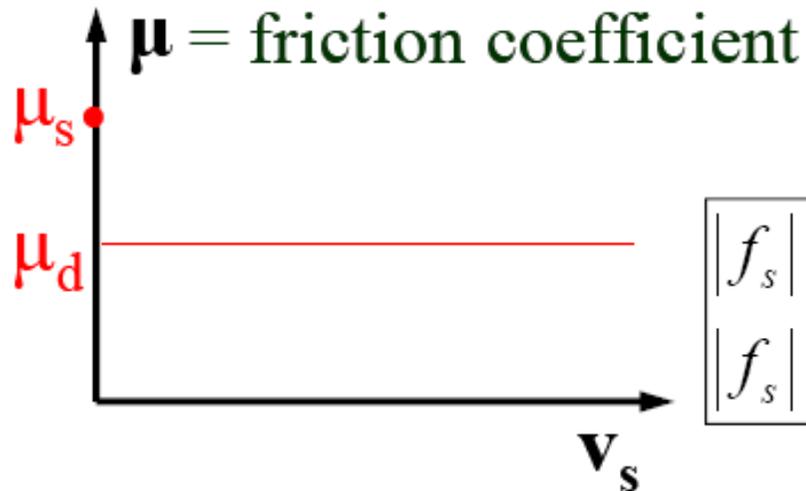
Where A depends on the elastic moduli and the particle radius.

$$F_n^{diss} = B \sqrt{\xi} \dot{\xi} \theta(\xi)$$

B determines r

Similar, more complicated expressions are valid for F_t

Coulomb friction with rigid particles:



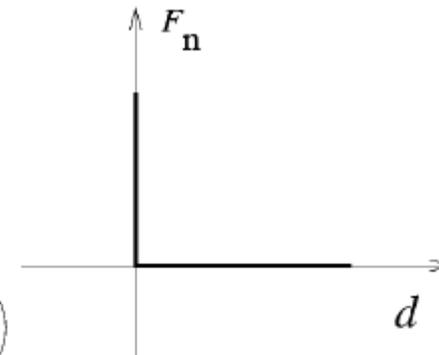
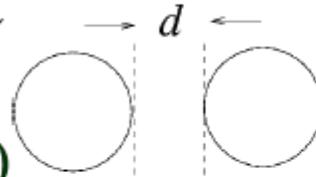
$ f_s \leq \mu_s f_n $	if	$v_s = 0$
$ f_s \leq \mu_d f_n $	if	$v_s \neq 0$

static case

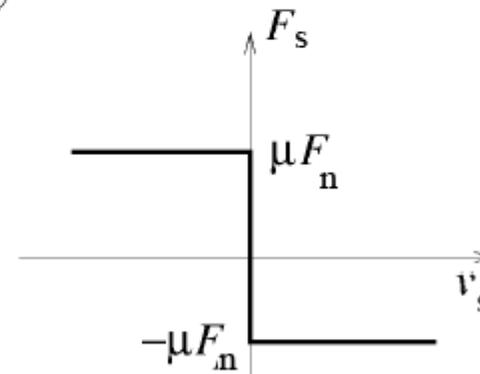
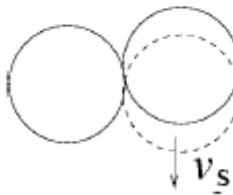
dynamic case

Non-smooth dynamics

Signorini-Graph: perfect volume exclusion (perfectly rigid particles)



Coulomb-Graph: friction between particles with relative tangential velocity v_s



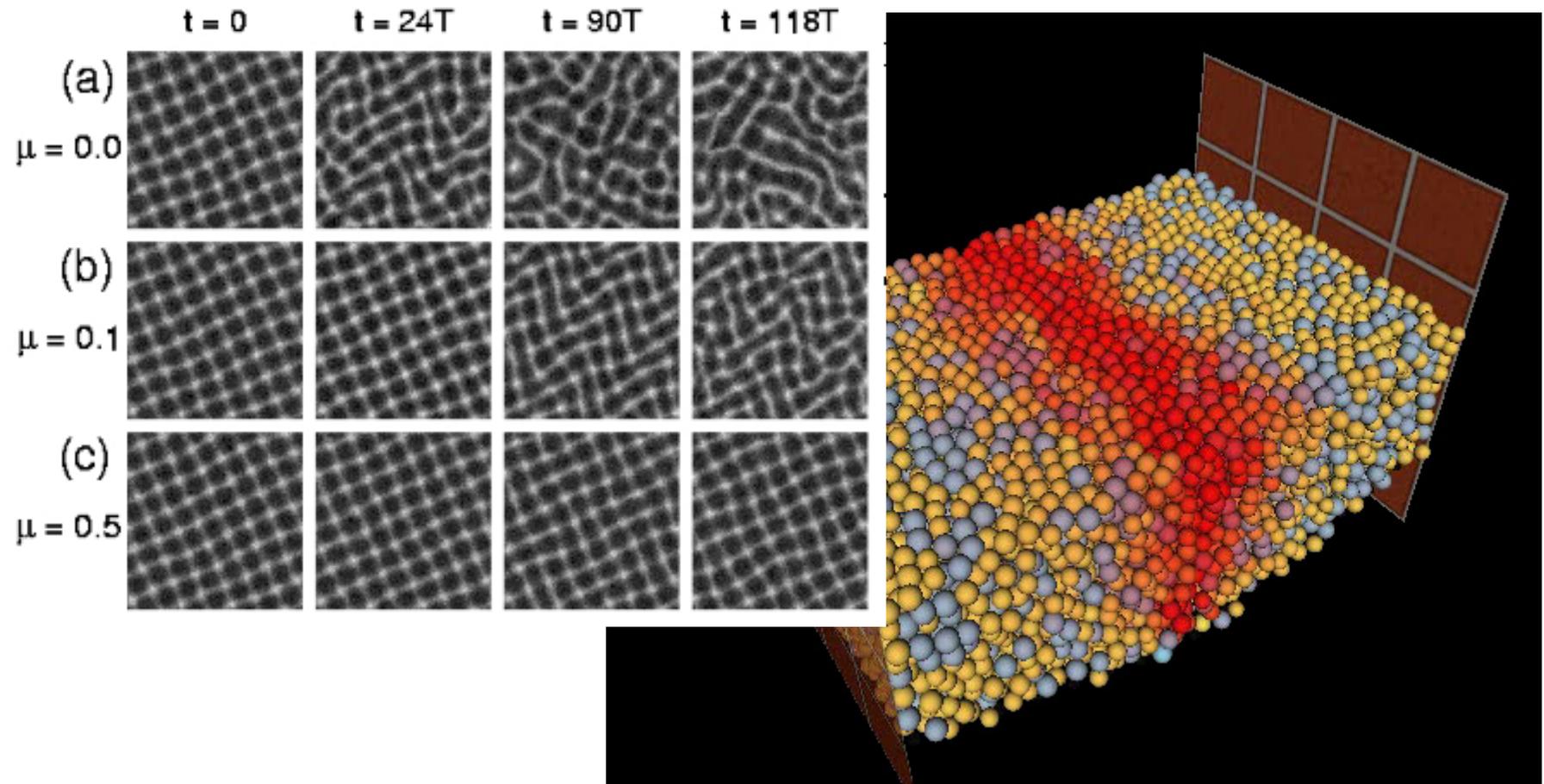
Contact dynamics

Shear

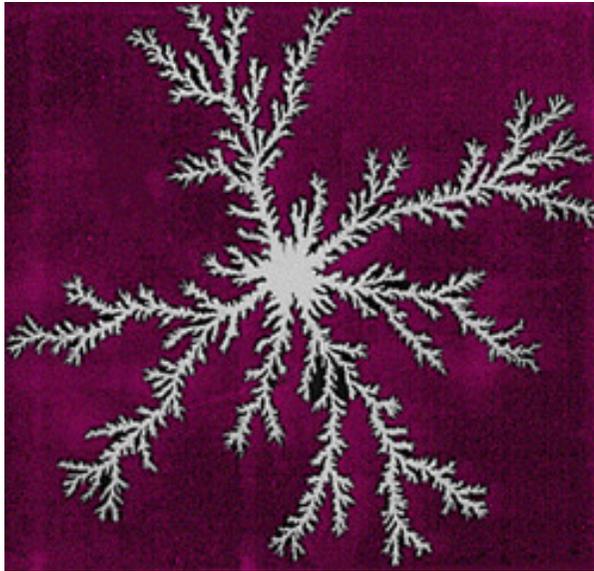
<http://www.ph.biu.ac.il/~rapaport/java-apps/grshear.html>

Vibrating plate

<http://www.ph.biu.ac.il/~rapaport/java-apps/grvib.html>



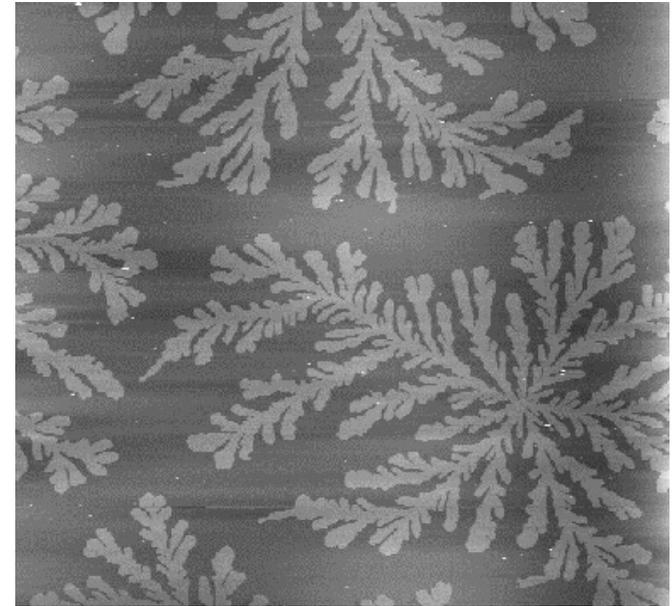
Algoritmikusan definiált modellek: Fraktál növekedés



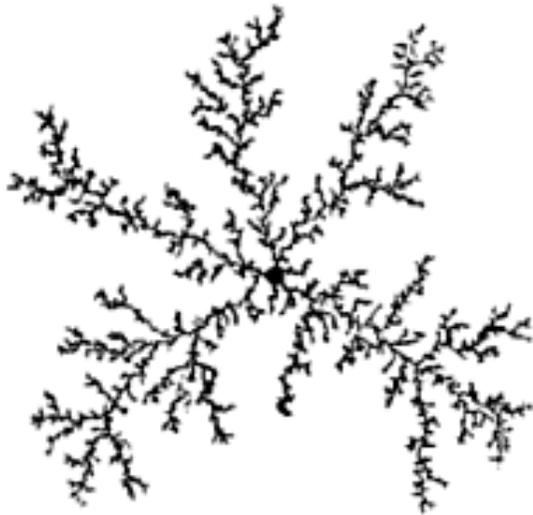
Electrochem. deposition



Mineralization

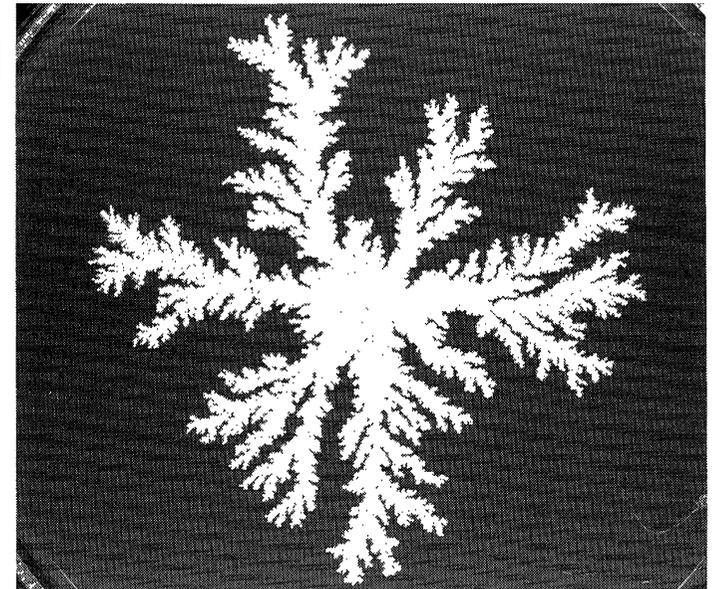


Surface crystallization



Disordered viscous fingering

Bacterial colony growth



Basic equations:

$$\nabla^2 u = 0 \quad u \text{ scalar field } (T, P, c \dots)$$

$$\mathbf{v}|_{\Gamma} = -C \nabla u|_{\Gamma} \quad \mathbf{v} \text{ velocity of the interface } \Gamma$$

$$u|_{\Gamma} = f(\nabla u, \kappa) \quad \kappa \text{ curvature (cutoff)}$$

+ disorder



Laplacian or gradient governed growth:

If there is a bump, the gradient increases (c.f. electrostatic peak effect) the bump grows... instability

+ screening:

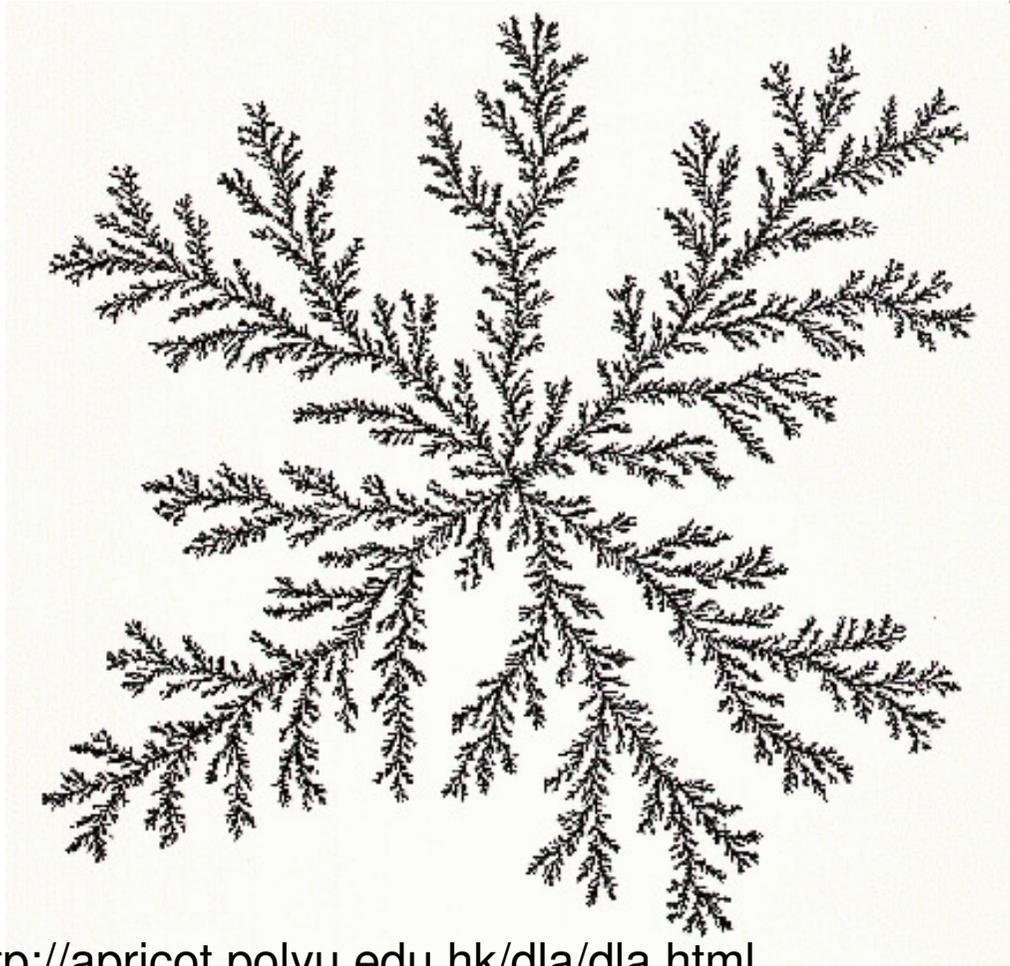
If 2 bumps grow, the faster will screen the slower and stop its growth

Simple model: Diffusion limited aggregation (DLA)

Start with a seed particle forming the initial aggregate.

* Another particle comes from infinity via a random walk until it sticks to the aggregate.

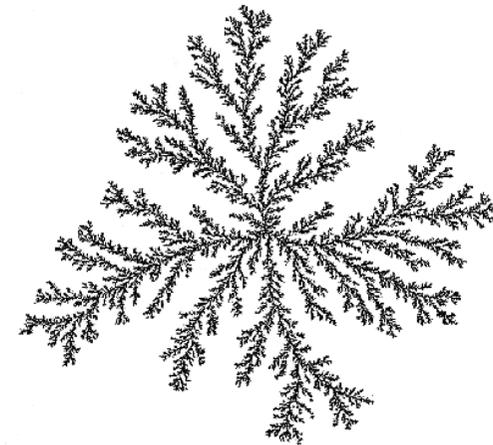
Goto *



100 million particles

Coarsened

Self-similar structure



1 million particles

In order to simulate (relatively) large samples tricks are needed

- No need to start from infinity: Birth ring surrounding the aggregate
- No need to let the particles walk far away: killing ring
- If far from the aggregate: large steps possible

For very large ($>10^7$) particles more tricks (fitted step size, dynamic storage)

Why are so terribly large aggregates needed?

Self similar fractals, scaling \rightarrow asymptotic behavior

How to measure fractal dimension?

Dimensions

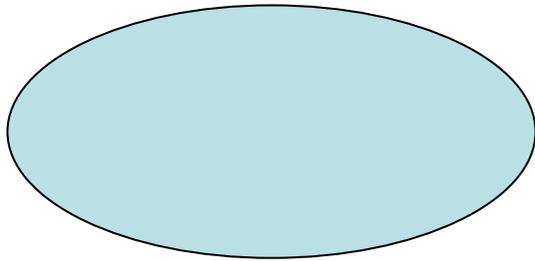
- Topological dimension:

Point: $d_t=0$, moving point: $d_t=1$, moving line, $d_t=2\dots$

- Embedding dimension:

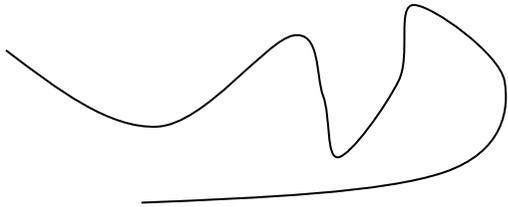
Number of independent directions

- Hausdorff (fractal) dimension



Area A is measured by covering the object with squares of size ℓ^2 . # of such boxes: N_ℓ .

$$A = \lim_{\ell \rightarrow 0} \ell^2 N_\ell$$



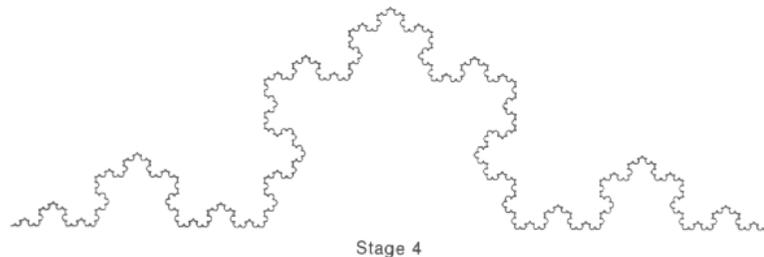
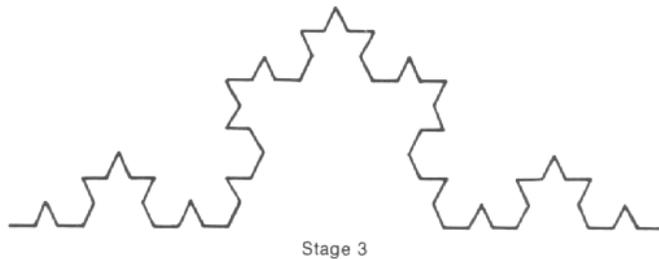
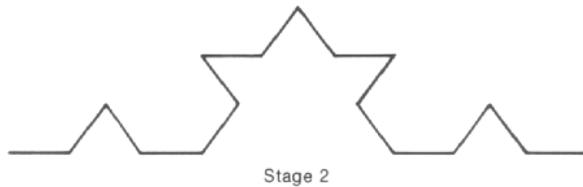
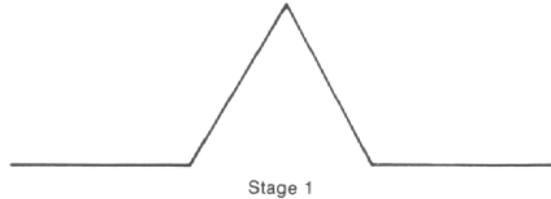
$$L = \lim_{\ell \rightarrow 0} \ell^1 N_\ell$$

In general:

$$M = \lim_{\ell \rightarrow 0} \ell^{d_t} N_\ell$$

M : mass

For a fractal this definition does not lead to a good result (0 or ∞)



Adapted from Benoit Mandelbrot, *Fractals*.

$$\ell = \left(\frac{1}{3}\right)^n; \quad N_\ell = 4^n; \quad d_t = 1$$

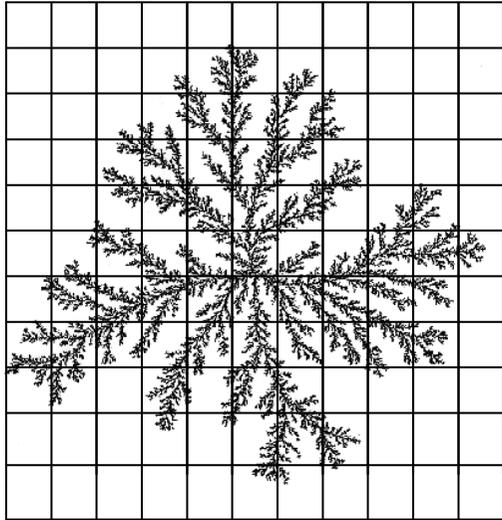
$$M = \lim_{n \rightarrow \infty} \left(\frac{4}{3}\right)^n = \infty$$

Instead of using d_t find the appropriate D fractal dimension such that

$$M = \lim_{\ell \rightarrow 0} \ell^D N_\ell \quad \text{is finite!}$$

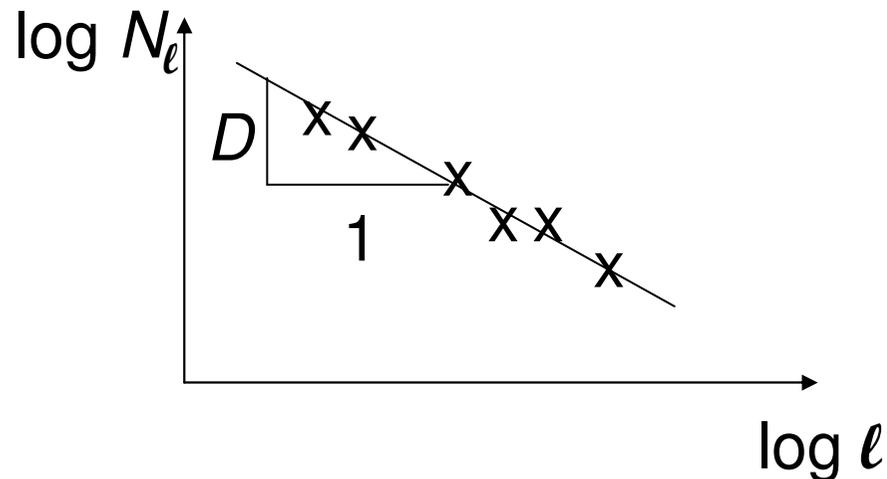
$$\text{Here: } D = \frac{\ln 4}{\ln 3} \quad (\text{Def of fractal: } d_t < D < d_e)$$

How to measure D for a random fractal?

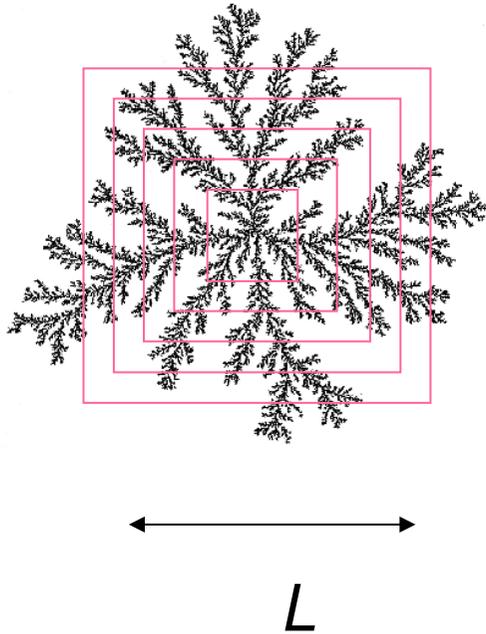


There is always a lower and an upper cutoff (e.g., particle size, radius of gyration).

1. Box counting: Use the definition of D . Cover the object with a mesh of mesh size ℓ , count the boxes where there is occupation. Plot log-log the dependence of N_ℓ vs ℓ .



2. Sand box method



$$M \sim L^D$$

3. Correlation functions

$$C(r) = \langle \rho(r) \rho(0) \rangle \sim r^{-\alpha}$$

$$\int C(r') d^d r' \sim r^{r-\alpha} \sim M(r)$$

$$D = d - \alpha$$