### Random Walk, Diffusion, and Cluster Growth

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### Outline

- 2D Random Walk
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- Diffusion
  - Mathematical background
  - Numerical results
- Cluster Growth
  - Background
  - Fractal Dimension



### 1D Random Walk

RMS distance of an ensemble of random walkers after n steps

$$\sqrt{\langle x_n^2 \rangle} = \sqrt{\sum_{i=1}^n \sum_{j=1}^n \langle \Delta x_i \Delta x_j \rangle} = \sqrt{n} \Delta x \tag{1}$$

Diffusive motion:

$$\langle x^2(t) \rangle = 2Dt \tag{2}$$

where  $D = v \cdot \Delta x/2 = (\Delta x)^2/(2\Delta t)$  is the diffusion constant.

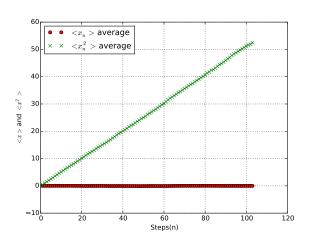


### 2D Random Walk (1)

Diffusive motion:

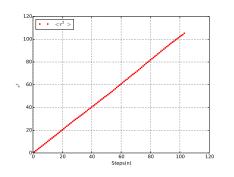
$$\langle r^2(t) \rangle = 2Dt \tag{3}$$

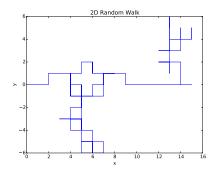
Where  $D = (\Delta x)^2/(4\Delta t)$ .





# 2D Random Walk (2)







### Diffusion

Diffusion equation

$$\frac{\partial \rho}{\partial t} = D\nabla^2 \rho \tag{4}$$

discretize  $t = k\Delta t, x = i\Delta x$ ,

$$\rho_{i,k+1} = \rho_{i,k} + D \frac{\Delta t}{\Delta x^2} (\rho_{i+1,k} + \rho_{i-1,k} - 2\rho_{i,k})$$
 (5)

with 
$$\Delta t < \frac{\Delta x^2}{2D}$$



### Normal Distribution

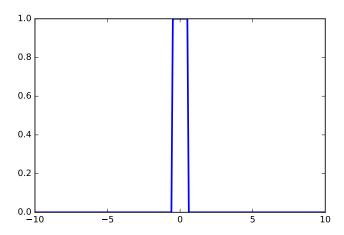
#### One-dimensional normal distribution

$$\rho(x,t) = \frac{1}{\sqrt{2\pi\,\sigma(t)^2}} \exp\left(-\frac{x^2}{2\sigma(t)^2}\right) \tag{6}$$

$$\langle x^2 \rangle = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma(t)} x^2 \exp(-\frac{x^2}{2\sigma(t)^2}) = \sigma(t)^2$$
 (7)

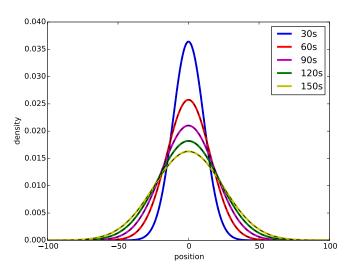


### Animation





### Diffusion time dependence



$$\langle x^2 \rangle = \sigma(t)^2 = 2Dt$$



# Cluster Growth with Diffusion Limited Aggregation (DLA)

#### DLA Algorithm:

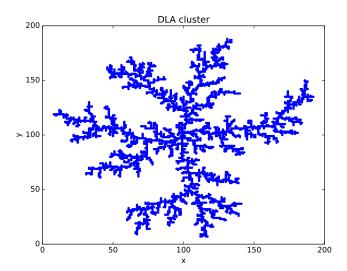
- Consider a lattice of points with a seed particle at the origin
- $oldsymbol{\circ}$  Release a particle from a random location a distance R from the origin
- Set the particle perform a random walk until it hits the perimeter of the cluster or strays too far
- If the particle hits the perimeter, add the particle to the cluster
- $\bullet$  Repeat until cluster reaches the desired size(Radius = 100)

We used a 2D Array for the lattice, and 0 for unoccupied lattice point, 1 for occupied lattice point.

Animation

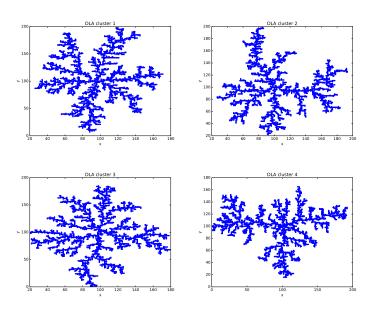


## Example DLA cluster





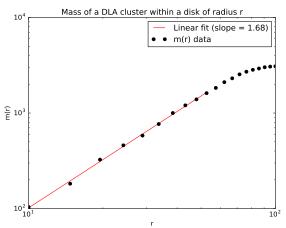
# Typical clusters





# Fractal Dimension $d_f$ (1)

$$m(r) \sim r^{d_f}$$
  
 $\log(m) = d_f \log(r)$ 





# Fractal Dimension $d_f$ (2)

Cluster	1	2	3	4	5	6	7	8	9	10
$d_f$	1.74	1.63	1.77	1.55	1.66	1.58	1.64	1.74	1.74	1.75

Average value over 10 runs:  $d_f = 1.68(07)$ 

Expected value:  $d_f = 1.65$ 

