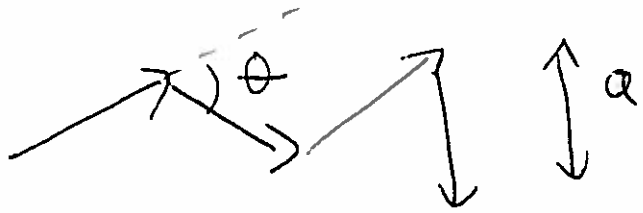


3. Root Mean Square Displacement X_{rms}



$|\vec{x}_i| = a$ Fixed step length

$\langle \vec{x}_i \rangle = 0$ no bias

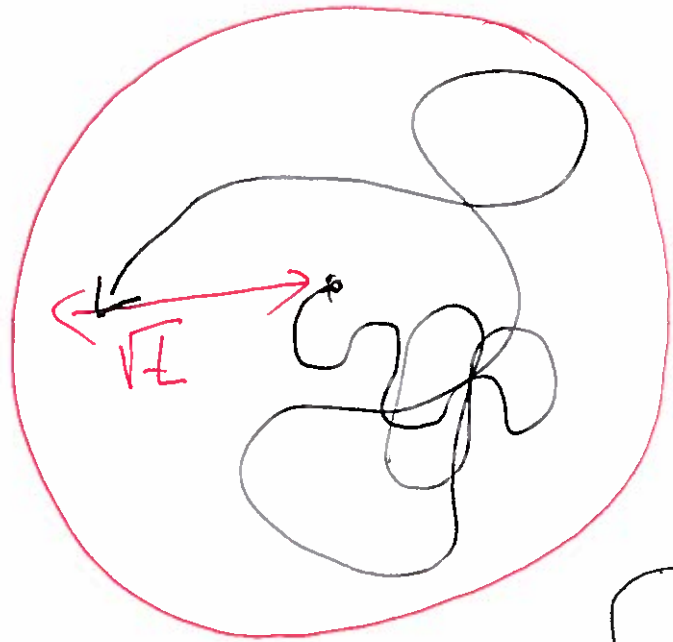
$\langle \vec{x}_i \cdot \vec{x}_j \rangle = 0$ no correlations
 $i \neq j$

$$\begin{aligned}\langle \vec{X}_N \rangle &= \left\langle \sum_{i=1}^N \vec{x}_i \right\rangle \\ &= 0\end{aligned}$$

$$\begin{aligned}\langle (\vec{X}_N)^2 \rangle &= \left\langle \left(\sum_{i=1}^N \vec{x}_i \right)^2 \right\rangle \\ &= \left\langle \sum_{i=1}^N \vec{x}_i^2 + \sum_{i \neq j} \vec{x}_i \cdot \vec{x}_j \right\rangle \\ &= Na^2\end{aligned}$$

$$\begin{aligned}X_{rms} &\equiv \sqrt{\langle \vec{X}_N^2 \rangle} \\ &= \sqrt{Na}\end{aligned}$$

4. Role of the Spatial Dimension d



density of visited sites

$$\rho \sim \frac{L}{(\sqrt{L})^d} \sim L^{1-d/2}$$

$$\rho \rightarrow \left\{ \begin{array}{l} 0 \\ \text{const. Int} \\ \infty \end{array} \right.$$

$$d > 2$$

transient
return uncertain

$$d = 2$$

$$d < 2$$

recurrent
return certain

5. Probability Distribution

$P(x,t)$ = probability that a random walk is at x at time t

$$= \frac{1}{2} P(x-1, t-1) + \frac{1}{2} P(x+1, t-1)$$

$$P(r,t) = \frac{t!}{r! l!} \left(\frac{1}{2}\right)^t$$

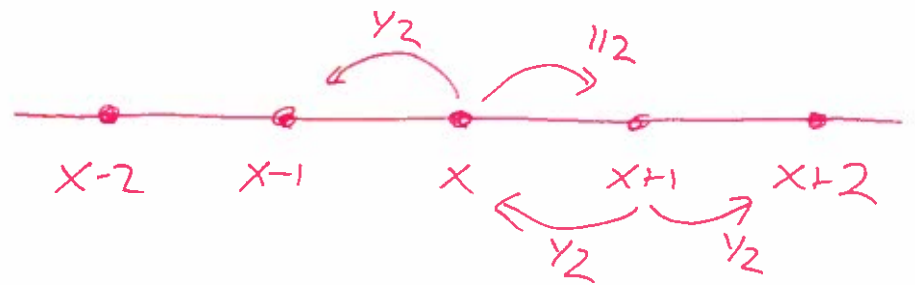
$$P(x,t) = \frac{t!}{\left(\frac{t+x}{2}\right)! \left(\frac{t-x}{2}\right)!} \left(\frac{1}{2}\right)^t$$

$\xrightarrow{t \rightarrow \infty}$

Stirling's approx

$$\sqrt{\frac{2}{\pi t}} e^{-x^2/2t}$$

Gaussian



r = # steps right
 l = # steps left

$$r+l = t$$

$$r-l = x$$

$$r = \frac{t+x}{2} \quad l = \frac{t-x}{2}$$

5 (cont) Diffusion Equation

$$P(x, t + dt) = \frac{1}{2} P(x - dx, t) + \frac{1}{2} P(x + dx, t)$$

$$\cancel{P(x, t)} + dt \frac{\partial P}{\partial t} + \dots = \frac{1}{2} \left[\cancel{P(x, t)} - dx \frac{\partial P}{\partial x} + \frac{1}{2} dx^2 \frac{\partial^2 P}{\partial x^2} + \dots \right] \\ \frac{1}{2} \left[\cancel{P(x, t)} + dx \frac{\partial P}{\partial x} + \frac{1}{2} dx^2 \frac{\partial^2 P}{\partial x^2} + \dots \right]$$

$$\frac{\partial P}{\partial t} = \underbrace{\frac{(dx)^2}{2dt}}_{\equiv D} \frac{\partial^2 P}{\partial x^2}$$

lim dx, dt

such that $\frac{dx^2}{2dt} = \text{const} \equiv D$ diffusion coefficient

diffusion eqn

general dimension

$$\frac{\partial P}{\partial t} = D \nabla^2 P$$

$$[D] = \frac{l^2}{t}$$

$$; \quad [x^2] = [Dt]$$

5 (cont.) Solution to the Diffusion Equation

$$\int_{-\infty}^{\infty} \left[\frac{\partial P}{\partial t} = D \frac{\partial^2 P}{\partial x^2} \right] e^{ikx} dx \quad P(x, t=0) = \delta(x)$$

$$\int_0^{\infty} \left[\frac{\partial P(k, t)}{\partial t} = -DK^2 P(k, t) \right] e^{-st} dt$$

$$s P(k, s) - \underbrace{P(k, t=0)} = -DK^2 P(k, s)$$

$$P(k, s) = \frac{1}{s + DK^2}$$

$$P(k, t) = \frac{1}{2\pi i} \int_C \frac{1}{s + DK^2} e^{st} ds$$

$$= e^{-DK^2 t}$$

$$P(x, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-DK^2 t} e^{-ikx} dk$$

Fourier transform

$$f(k, t) = \int_{-\infty}^{\infty} f(x, t) e^{ikx} dx$$

$$f(x, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} f(k, t) e^{-ikx} dk$$

Laplace transform

$$f(x, s) = \int_0^{\infty} f(x, t) e^{-st} dt$$

$$f(x, t) = \frac{1}{2\pi i} \int_C f(x, s) e^{st} ds$$

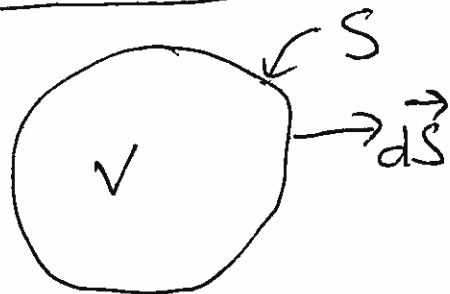
$$P(x, t) = \frac{1}{\sqrt{4\pi Dt}} e^{-x^2/4Dt} \quad d=1$$

$$P(\vec{r}, t) = \frac{1}{(4\pi Dt)^{d/2}} e^{-r^2/4Dt} \quad d \text{ arbitrary}$$

Gaussian

5. (cont) Connection with Kinetic Theory and Particle Conservation

Conservation

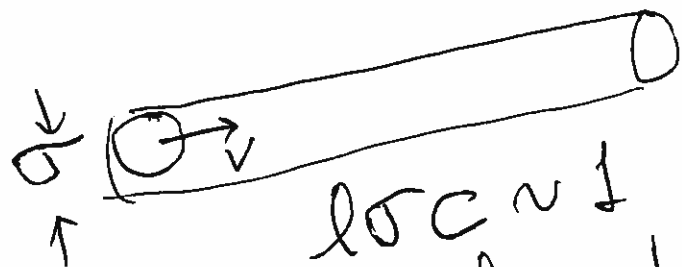


$$\frac{\partial N}{\partial t} = - \int \vec{j} \cdot d\vec{S}$$

$$\int \frac{\partial c(r,t)}{\partial t} dV = - \int \vec{\nabla} \cdot \vec{j} dV$$

$$\boxed{\frac{\partial c}{\partial t} = - \vec{\nabla} \cdot \vec{j}}$$

mean-free path

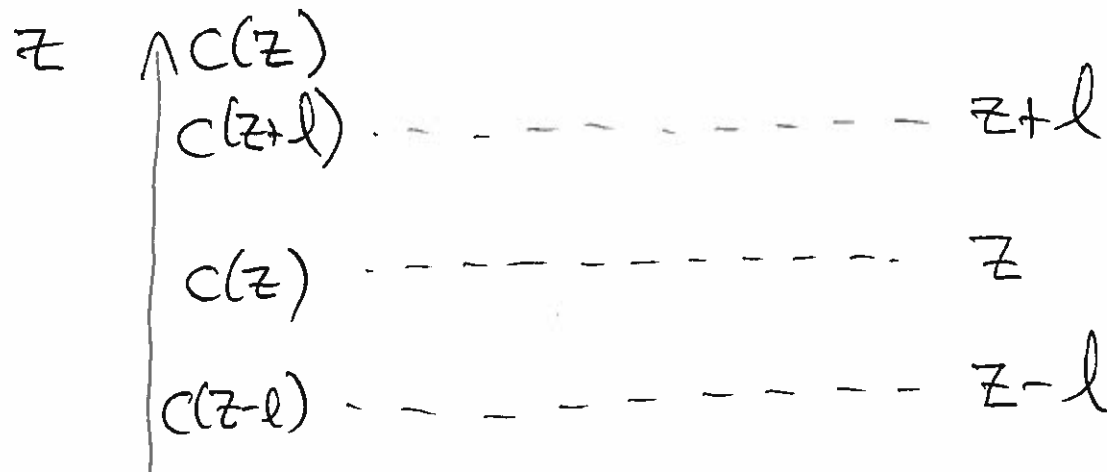


$$l \sigma c \sim 1$$

$$\Rightarrow l \sim \frac{1}{c \sigma}$$

collision

determine j by kinetic



$$j_{up} \approx c(z-l) \frac{v}{6}, \quad j_{down} = c(z+l) \frac{v}{6}$$

$$j_{net} = \frac{v}{6} [c(z-l) - c(z+l)] \approx -\frac{vl}{3} \frac{\partial c}{\partial z}$$

$$\boxed{j \equiv -D \frac{\partial c}{\partial z}}$$

$$\boxed{D \approx vl}$$

$$\boxed{\frac{\partial c}{\partial t} = - \vec{\nabla} \cdot \vec{j} = D \nabla^2 c}$$

$$J = -\frac{v l}{3} \frac{\partial c}{\partial x} = -\boxed{D} \frac{\partial c}{\partial x}$$

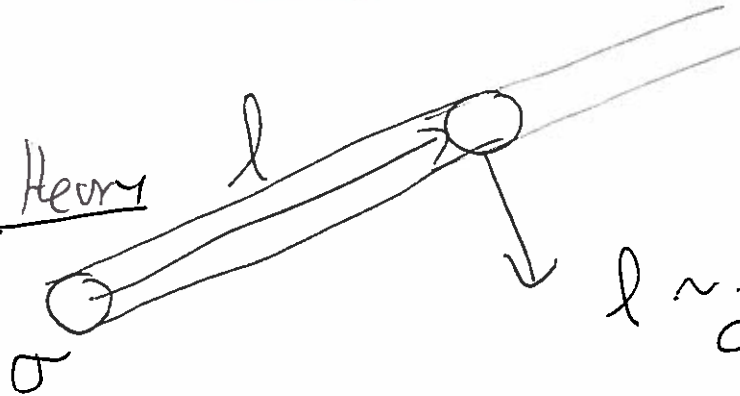
dissipation

$$\frac{\partial c}{\partial t} = \frac{(dx)^2}{2dt} \frac{\partial^2 c}{\partial x^2} = \boxed{D} \frac{\partial^2 c}{\partial x^2}$$

equal. — are these equal?

Fluctuation yes: fluctuation dissipation relation

From kinetic theory



$cl\sigma \sim 1$ defines collision

$$l \sim \frac{1}{c\sigma} \sim \frac{1}{10^{20}/\text{cm} \cdot 10^{-5}\text{cm}} \sim 10^{-5}\text{cm}$$

$\sim 100-1000$
molecular
diameters

$$\tau \sim \frac{l}{v} \sim \frac{10^{-5}\text{cm}}{30000\text{cm/sec}}$$

$$\sim 10^{-9} - 10^{-10} \text{ sec}$$

RW picture

$$\langle X^2 \rangle = N l^2$$

$$= \frac{t}{\tau} l^2 = \frac{t}{l/v} l^2 = \underbrace{(vl)}_D t$$

6. Central Limit Theorem

$x, p(x)$

Assume are i.i.d variables
(independent identically distributed)

Assume $\langle x \rangle < \infty$
 $\langle x^2 \rangle < \infty$

$$\text{Then } P_N(X) = \frac{1}{\sqrt{2\pi N\sigma^2}} e^{-\frac{(X - N\langle x \rangle)^2}{2N\sigma^2}}$$

Central
Limit
Theorem

$$\sigma^2 = \langle x^2 \rangle - \langle x \rangle^2$$

$$\langle X \rangle = \int X P_N(x) dx = N \langle x \rangle$$

$$\begin{aligned} \langle X^2 \rangle - \langle X \rangle^2 &= \int X^2 P_N(x) dx - \langle X \rangle^2 \\ &= N \sigma^2 \end{aligned}$$

6. Proof of Central Limit Theorem

$$P_N(x) = \int P_{N-1}(x') p(x' \rightarrow x) dx'$$

Fourier transform $P_N(k) = P_{N-1}(k) p(k) = P_{N-2}(k) (p(k))^2$
 $= P_0(k) [p(k)]^N = [p(k)]^N$

$$P_N(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} [p(k)]^N e^{ikx} dk$$

$$p(k) = \int p(x) e^{ikx} dx = \int p(x) \left[1 + ikx - \frac{k^2 x^2}{2} + \dots \right] dx$$
$$= 1 + i k \cancel{\langle x \rangle} - \frac{k^2}{2} \langle x^2 \rangle + \dots$$

$$P_N(x) = \frac{1}{2\pi} \int \left[1 - \frac{k^2 \langle x^2 \rangle}{2} \right]^N e^{-ikx} dk \approx \frac{1}{2\pi} \int e^{-Nk^2 \langle x^2 \rangle / 2 - ikx} dk$$
$$= \frac{1}{\sqrt{2\pi N \langle x^2 \rangle}} e^{-x^2 / 2N \langle x^2 \rangle}$$

- Gaussian
- independent of almost all details of the individual steps

(c) Cont) Failure of the Central Limit Theorem

$$p(x) = \begin{cases} 0 & x < 1 \\ \mu x^{-(1+\mu)} & x > 1 \end{cases} \quad \langle x \rangle = \infty$$

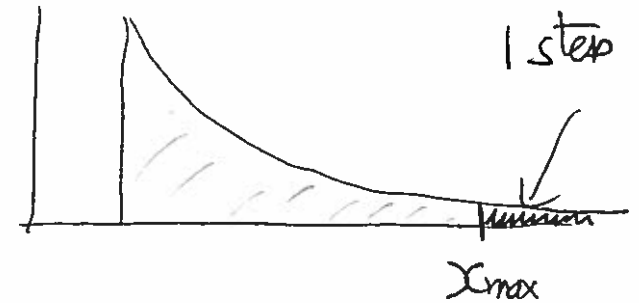
$$\mu < 1$$

Consider N steps

x_{\max} determined by

$$\int_{x_{\max}}^{\infty} p(x) dx = \frac{1}{N}$$

$$x_{\max} \sim N^{1/\mu}$$



$$p_{\text{eff}} = \begin{cases} 0 & x < 1 \\ \frac{\mu x^{-(1+\mu)}}{1 - x_{\max}^{-\mu}} & 1 < x < x_{\max} \\ 0 & x > x_{\max} \end{cases}$$

$$\langle x \rangle_{\text{eff}} = \frac{\mu}{1 - x_{\max}^{-\mu}} \int_1^{x_{\max}} x x^{-(1+\mu)} dx = \begin{cases} \text{finite} & \mu > 1 \\ \ln x_{\max} \sim \ln N & \mu = 1 \\ x_{\max}^{1-\mu} \sim N^{\frac{1}{\mu}-1} & \mu < 1 \end{cases}$$

$$\langle X \rangle_N = N \langle x \rangle_{\text{eff}}$$

$$= \begin{cases} N & \mu > 1 \\ N \ln N & \mu = 1 \\ N \cdot N^{\frac{1}{\mu}-1} \sim N^{\frac{1}{\mu}} & \mu < 1 \end{cases}$$

$$x_{\max} \sim N^{\frac{1}{\mu}}$$

7. First Passage Phenomena

Questions: 1. What is the prob of eventually hitting 0 when starting from x ?
2. What is the time to hit the origin?

Solve
$$\begin{cases} \frac{\partial C}{\partial t} = D \frac{\partial^2 C}{\partial x^2} \\ C(x, t=0) = \delta(x-x_0) \\ C(x=0, t) = 0 \quad \text{absorbing} \end{cases}$$

$$C(x, t) = \frac{1}{\sqrt{4\pi Dt}} \left[e^{-(x-x_0)^2/4Dt} - e^{-(x+x_0)^2/4Dt} \right]$$

$$J = -D \frac{\partial C}{\partial x} \Big|_{x=0} = \frac{D}{\sqrt{4\pi Dt}} \left[\frac{x-x_0}{2Dt} - \frac{(x+x_0)}{2Dt} \right] e^{-x_0^2/4Dt}$$

$$j = -\frac{x_0}{\sqrt{4\pi Dt^3}} e^{-x_0^2/4Dt} \equiv \text{first passage prob to 0}$$

$$\bar{J} = \int_0^\infty j dt = 1$$

return is certain

$$\langle t \rangle = \int_0^\infty t j(t) dt / \int_0^\infty j(t) dt = \infty$$

return time = ∞

7. Application: First Passage in an Interval



$E(x)$ = prob that I win starting with x
 $t(x)$ = time of the game starting with x

$$E(x) = \sum_{\text{paths}} P_{x \rightarrow L} = \frac{1}{2} \sum_{\text{paths}'} P_{x+dx \rightarrow L} + \frac{1}{2} \sum_{\text{paths}''} P_{x-dx \rightarrow L}$$

$$= \frac{1}{2} E(x+dx) + \frac{1}{2} E(x-dx) \quad \begin{matrix} E(0) = 0 \\ E(L) = 1 \end{matrix}$$

$$E''(x) = 0 \rightarrow E(x) = x/L$$

even simpler $\langle x \rangle \equiv \text{invariant}$

$$\begin{matrix} t=0 & \langle x \rangle = x \\ t=\infty & \langle x \rangle = (1-E) \cdot 0 + E \cdot L \end{matrix} \quad E(x) = x/L$$

$$t(x) = \sum_{\text{paths}} t_{x \rightarrow 0, L} = \frac{1}{2} (dt + \sum_{\text{paths}'} t_{x+dx \rightarrow 0, L}) + \frac{1}{2} (dt + \sum_{\text{paths}''} t_{x-dx \rightarrow 0, L})$$

$$t(x) = dt + \frac{1}{2} t(x+dx) + \frac{1}{2} t(x-dx)$$

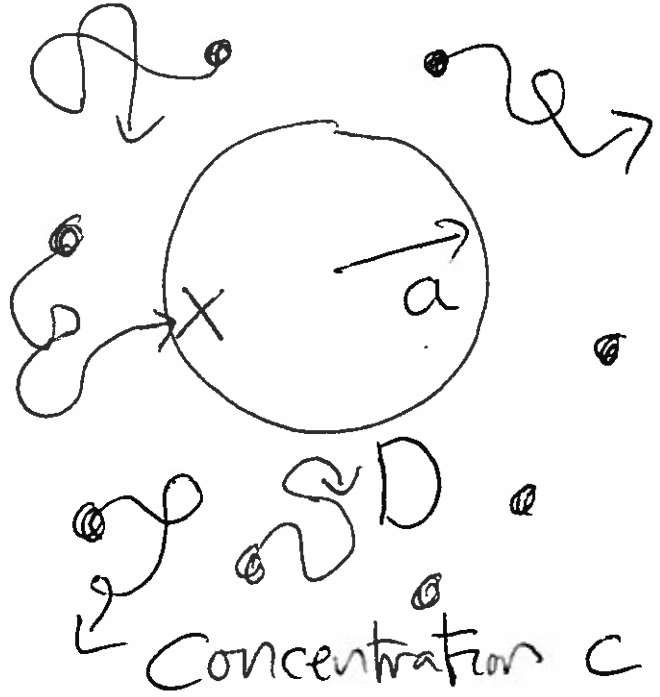
$$0 = dt + \frac{1}{2} (dx)^2 t''(x) \Rightarrow t''(x) = -\frac{2dt}{(dx)^2} = -\frac{1}{D}$$

$$Dt'' = -1 \quad t(0) = t(L) = 0$$

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



7. Application: Reaction Rate Theory



$k \equiv$ reaction rate

$= \frac{\text{\# particles absorbed / time}}{c}$

$$[k] = \frac{l^d}{t} ; k = k(D, a) \quad \nearrow [D] = l^2/t$$

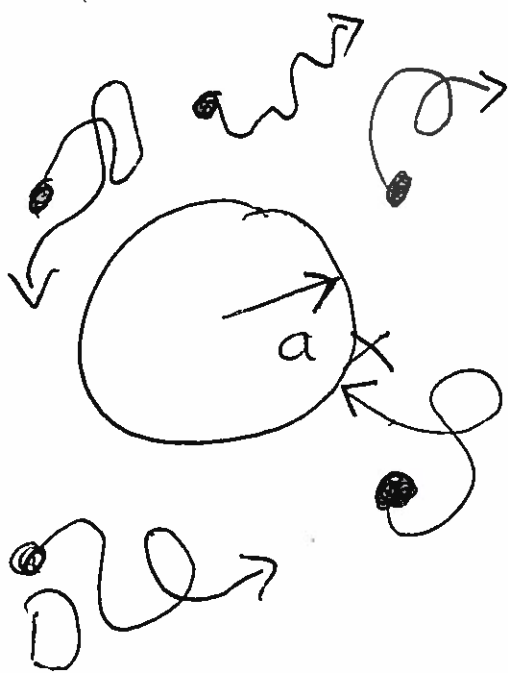
$$\Rightarrow k \propto Da^{d-2}$$

$$d=3: k \propto a !!$$

$$d < 2: k \uparrow \text{ as } a \downarrow !$$

\Rightarrow new dependence ~~of~~ on system parameters

7. Application: Reaction Rate Theory in 3 Dimensions



Solve

$$\begin{cases} \frac{\partial c}{\partial t} = D \nabla^2 c \\ c(r > a, t=0) = 1 \\ c(r=a, t > 0) = 0 \end{cases}$$

Instead

$$\begin{cases} D \nabla^2 c = 0 \\ c(r=a) = 0 \\ c(r \rightarrow \infty) = 1 \end{cases}$$

$$C(r) = 1 - \frac{a}{r} \quad \equiv \text{escape probability}$$

$$K = \int (-D \vec{\nabla} c) \cdot (-d\vec{S})$$

$$= Da \int \frac{1}{a^2} dS$$

$$= 4\pi Da \left(1 + \frac{a}{\sqrt{\pi Dt}} \right)$$