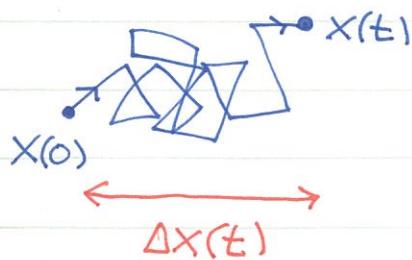


HH0049 Einstein relation in Brownian Motion.

Einstein was the first person to develop a quantitative theory for Brownian motions. Later, Perrin performed detail expt to verify the theory and won the Nobel Prize in 1926 ☺.



Later, we will derive the important relation

$$(\Delta x)^2 = 2Dt$$

where $(\Delta x)^2 = \langle x_i^2(t) \rangle$
since $\langle x_i(t) \rangle = 0$.

∅ Diffusion from fluctuations: Starting from the definition, $(\Delta x)^2 = \langle x^2(t) \rangle = \int_0^t dt_1 \int_0^{t_1} dt_2 \langle v(t_1)v(t_2) \rangle$

We're given $(\Delta x)^2 = 2Dt$ in the long-time limit,

$$D = \lim_{t \rightarrow \infty} \frac{1}{2} \frac{d(\Delta x)^2}{dt} = \lim_{t \rightarrow \infty} \int_0^t dt_2 \langle v(t)v(t_2) \rangle \quad \begin{matrix} \text{velocity} \\ \text{correlation fn.} \end{matrix}$$

Making use of translational invariance in time,

$$\langle v(t)v(t_2) \rangle = \langle v(t-t_2)v(0) \rangle \text{ and change variable } t' = t-t_2$$

$$\rightarrow D = \lim_{t \rightarrow \infty} \int_0^t dt' \langle v(t')v(0) \rangle$$

long-time limit
expression for D ??

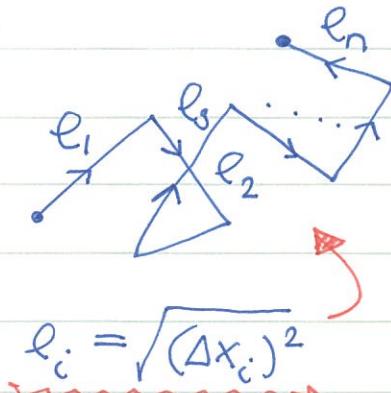
Within relaxation time approximation $\langle v(t')v(0) \rangle = \langle v^2 \rangle e^{-t'/\tau_c}$

$$D = \int_0^\infty dt' \langle v^2 \rangle \cdot e^{-t'/\tau_c} = \langle v^2 \rangle \cdot \tau_c = \frac{\tau_c}{M} \quad \text{YES} \Sigma$$

∅ Fractal dimension: Consider the total length of the path at time t with observation time interval τ .

The path is separated into $n = t/\tau$ segments

$$L_\tau = l_1 + l_2 + \dots + l_n$$



$$l_i = \sqrt{(\Delta x_i)^2}$$

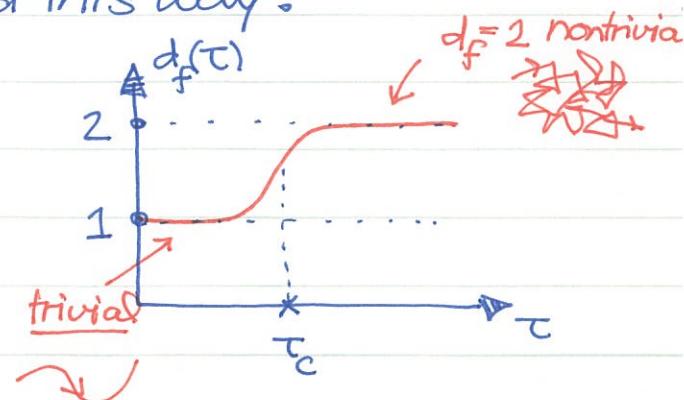
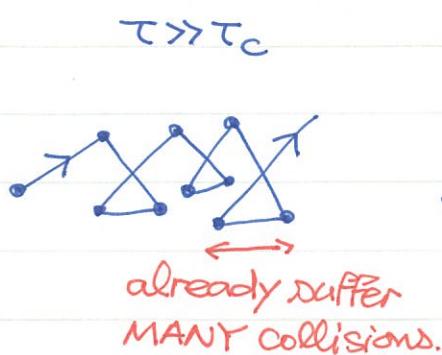
For $\tau \gg \tau_c$, each segment is described by $(\Delta x_i)^2 = 2D\tau$

$$L_\tau = \sum_{i=1}^n \sqrt{(\Delta x_i)^2} = n \cdot \sqrt{2D\tau} = [\sqrt{2D} \cdot t] \cdot \frac{1}{\sqrt{\tau}} \propto \tau^{1-d_f}$$

Thus, the fractal dimension $d_f = 2$. On the other hand, for $\tau \ll \tau_c$, the motion is ballistic $\ell_i \approx \bar{c} \cdot \tau$.

$$L_\tau = \sum_{i=1}^n \ell_i \approx n \bar{c} \cdot \tau = \bar{c} \tau \rightarrow d_f = 1 \text{ as ordinary curve.}$$

The difference can be understood this way:

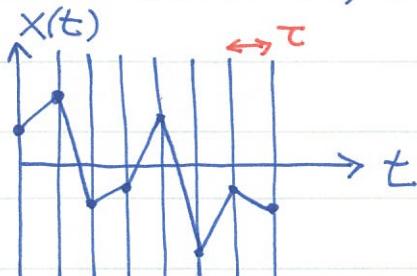


① **Probability distribution:** Now we want to derive the probability distribution function at microscopic level. Introduce $P(x, t)$:

$\frac{dx}{x \ x+dx}$

$P(x)dx = \text{prob. to find particle in } dx$

For convenience, let's make time discrete: $t = n\tau$ will take $\tau \rightarrow 0$ later.



Write down the master equation —

$$P(x, t+\tau) = \int_{-\infty}^{+\infty} d\Delta f_\tau(\Delta) P(x-\Delta, t),$$

where $f_\tau(\Delta)$ is the probability density for the particle to move Δ within the time interval τ .

$$\rightarrow \int_{-\infty}^{+\infty} d\Delta f_\tau(\Delta) = 1 \text{ and } f_\tau(\Delta) = f_\tau(-\Delta) \quad \text{no intrinsic drift.}$$

Now we are ready to expand the master equation ...

$$\cancel{P + \frac{\partial P}{\partial t} \cdot \tau \approx \int_{-\infty}^{+\infty} d\Delta f_\tau(\Delta) \left[P - \Delta \frac{\partial P}{\partial x} + \frac{1}{2} \Delta^2 \frac{\partial^2 P}{\partial x^2} \right]}$$

$$= \cancel{P} - \langle \Delta \rangle \frac{\partial P}{\partial x} + \frac{1}{2} \langle \Delta^2 \rangle \frac{\partial^2 P}{\partial x^2}$$

The master equation is greatly simplified if the higher order terms can be ignored,

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

with

$$D = \lim_{\tau \rightarrow 0} \frac{1}{2\tau} \int_{-\infty}^{+\infty} d\Delta f_\tau(\Delta) \Delta^2$$

diffusion equation.

short-time expression for D

Suppose $P(x, t=0) = \delta(x)$ → particle at the origin

In the Fourier space, $P(k, t=0) = \int dx e^{-ikx} \delta(x) = 1$.

The Diffusion eq. is

$$\frac{\partial P(k, t)}{\partial t} = -Dk^2 P(k, t)$$

simple to solve.

The solution is

$$P(k, t) = P(k, 0) e^{-Dk^2 t} = e^{-Dk^2 t} \quad \text{just Fourier transform it}$$

back to the coordinate space →

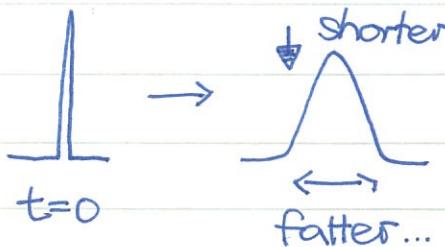
$$P(x, t) = \frac{1}{\sqrt{4\pi Dt}} e^{-\frac{x^2}{4Dt}}$$

The probability distribution for the Brownian motion is just the normal distribution. It is straightforward to compute the variance $\langle x^2 \rangle$.

$$\langle x^2 \rangle = 2Dt \rightarrow$$

$$(\Delta x)^2 = 2Dt$$

diffusion relation claimed previously.

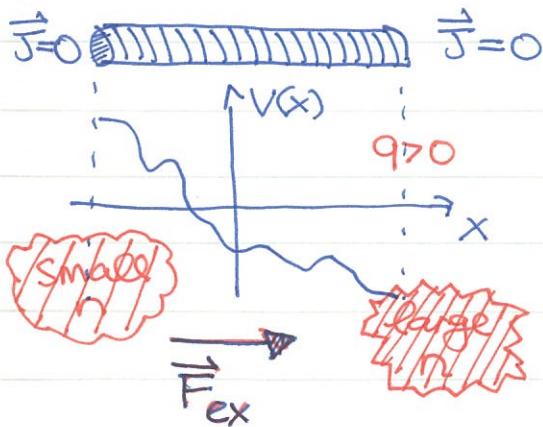


$$\Delta x = \sqrt{2Dt}$$

Q: Why is $P(x, t)$ a normal distribution?

① Einstein relation, at last... Now we would like to relate the diffusion constant D and the mobility $\tilde{\mu}$ again.

Consider an isolated wire with no current in or out. After thermal equilibrium is reached, the density dist.



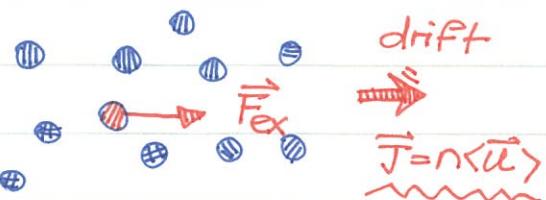
$$n(\vec{r}) = n_0 e^{-qV(\vec{r})/\tau}$$

and the total current density is zero everywhere. There are two contributions to \vec{J}_n :

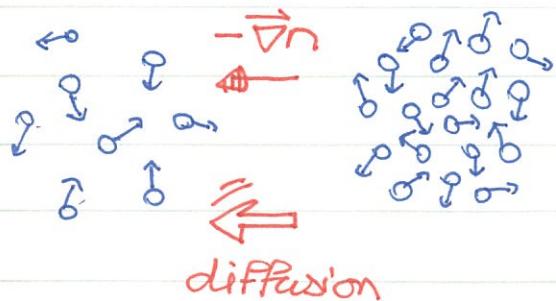
- ① external force $\vec{F}_{ex} = q\vec{E}$
- ② density gradient $\vec{\nabla}n$.

The drift velocity is $\langle \vec{u} \rangle = \tilde{\mu} \vec{E} = -\tilde{\mu} \vec{\nabla}V(\vec{r})$ giving rise to a current density

$$\vec{J}_n(\text{drift}) = n \langle \vec{u} \rangle = -n \tilde{\mu} \vec{\nabla}V$$



On the other hand, the density gradient gives rise to counter propagating current, as described by Fick's law:



law:

$$\vec{J}_n(\text{diffusion}) = -D \vec{\nabla}n$$

$$= D \cdot \left(\frac{q}{\tau} \vec{\nabla}V \right) \cdot n$$

Combine both contributions together,

$$(\text{drift}) + (\text{diffusion}) = 0 \rightarrow -\tilde{\mu} n \vec{\nabla}V + \frac{Dq}{\tau} n \vec{\nabla}V = 0$$

The detail balance for current leads to the Einstein relation,

$$D = \frac{\tilde{\mu}}{q} \cdot \tau$$

In some books, mobility is defined as $\langle \vec{u} \rangle = \tilde{\mu} \vec{F}_{ex}$, Einstein relation becomes $D = \tilde{\mu} \tau$

① Brownian motion in stock market: Consider the index of stock market $S(t)$ where $t = 0, \tau, 2\tau, \dots$. Introduce the (growth) ratio R_i defined as $R_1 = \frac{S(\tau)}{S(0)}, R_2 = \frac{S(2\tau)}{S(\tau)}, \dots$

At time $t=n\tau$, the index of the stock market is

$$S(n\tau) = \frac{S(n\tau)}{S((n-1)\tau)} \cdot \frac{S((n-1)\tau)}{S((n-2)\tau)} \cdots \frac{S(2\tau)}{S(\tau)} \cdot \frac{S(\tau)}{S(0)} \cdot S(0)$$

$$= \underbrace{R_n R_{n-1} \cdots R_2 R_1}_{\text{positive factors}} \cdot S(0), \text{ note that } 0 < R_i < \infty \text{ are}$$

Introduce the random variable $X \equiv \log [S(n\tau)/S(0)]$.

$$X = \sum_{i=1}^n \log R_i = x_1 + x_2 + \cdots + x_n \quad \text{where } x_i = \log R_i$$

According to central limit theorem, the probability dist. for X is normal distribution if x_i are independent with the same (though unknown) distribution.

$$P(x, n) = \frac{1}{\sqrt{2\pi\sigma^2 n}} e^{-\frac{(x-n\mu)^2}{2\sigma^2 n}}$$

$$\langle (x_i)^2 \rangle = \sigma^2$$

$$\langle x_i \rangle = \mu$$

Compare with diffusion, $\sigma^2 n = 2DT$

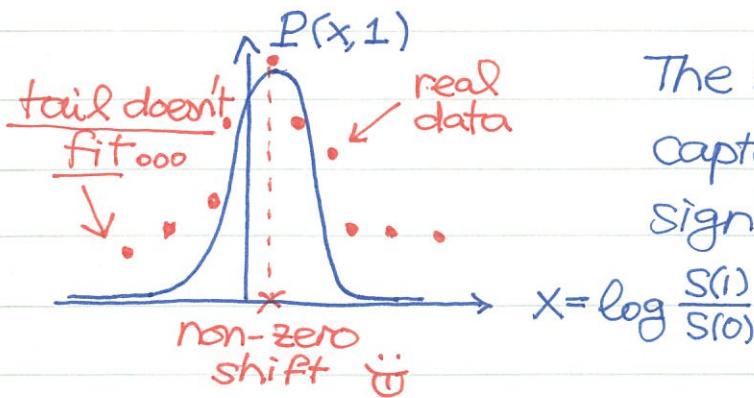
$$\rightarrow D = \sigma^2 / 2\tau$$

The drift velocity $n\mu = \langle u \rangle t \rightarrow \langle u \rangle = \mu / \tau$

Rewrite the dist. fn as

$$P(x, t) = \frac{1}{\sqrt{4\pi D t}} e^{-\frac{(x-\langle u \rangle t)^2}{4Dt}}$$

$\log[S(t)/S(0)]$ moves like Brownian motion !!



The simple model more or less captures the main trend with significant tail corrections.



2012.0520

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