

Elementary Physics of Random Walks and Diffusion

Basic ideas:

- random walk \rightarrow stochastic \rightarrow evolution of mean square
- Markov Process \rightarrow no memory, step-to-step
Each step uncorrelated and unbiased,
Each set by microscopic pdf.

Diffusion equation:

$$\frac{\partial n}{\partial t} + \nabla \cdot \Gamma = S_{c,0}$$

conserving flux

Fick's Law

$$\Gamma = -D \nabla n \rightarrow \text{where from?}$$

Generally: seek macro-density evolution from micro-step probability

i.e.

$n(x,t)$ evolution from:

transition probability
of step ΔX in Δt

2.

$$n(x, t + \Delta t) = \int d(\Delta X) \left[T(\Delta X, \Delta t) n(x - \Delta X, t) \right]$$

density up-dated
to Δt ahead

Chapman
- Kolmogorov Eqn.

density
one step
away

n evolves by small random kicks, so

$$n(x, t) + \Delta t \frac{\partial n}{\partial t} = \int d(\Delta X) T(\Delta X, \Delta t) \left[n(x, t) \right.$$

$$\left. - \Delta X \frac{\partial n}{\partial x} + \frac{(\Delta X)^2}{2} \frac{\partial^2 n}{\partial x^2} \right]$$

$$\int d(\Delta X) T = 1 \quad (\text{probability normalizable})$$

$$\int d(\Delta X) \Delta X T = \langle \Delta X \rangle \rightarrow \text{mean step}$$

$$\int d(\Delta X) (\Delta X)^2 T = \langle \Delta X^2 \rangle \rightarrow \text{mean square step}$$

$$n(x, t) + \Delta t \frac{\partial n}{\partial t} = n(x, t) - \langle \Delta X \rangle \frac{\partial n}{\partial x} + \frac{\langle \Delta X^2 \rangle}{2} \frac{\partial^2 n}{\partial x^2}$$

10

$$\frac{\partial n}{\partial t} = - \frac{\partial}{\partial x} \left[\frac{\langle \Delta x \rangle}{\Delta t} n - \frac{\langle \Delta x^2 \rangle}{2\Delta t} \frac{\partial n}{\partial x} \right]$$

Important to note:

$\Delta x \rightarrow$ step size
 $\Delta t \rightarrow$ step time



Every random walk characterized by these.

Now: $D = \frac{\langle (\Delta x)^2 \rangle}{2\Delta t} \rightarrow$ diffusion coefficient

$V =$ drift speed $= \frac{\langle \Delta x \rangle}{\Delta t}$

(N.B. Obv. scheme not limited to space)

$$\Rightarrow \frac{\partial n}{\partial t} = - \frac{\partial}{\partial x} \left[V n - D \frac{\partial n}{\partial x} \right]$$

\rightarrow Fokker-Planck Eqn.

Important Points:

- $\langle (\Delta x)^2 \rangle^{1/2} < L$ assumed in expansion
 i.e. small kicks! (Boltzmann not so limited)

n should be regarded as coarse grained,
on small scale, i.e.
 $\Lambda \rightarrow \langle n \rangle$

- formulation of F.P.E., diffn requires
 $\int d(\Delta x) (\Delta x)^2 T < \infty$.

i.e. second moment of T must converge.

e.g. $T \sim \exp[-(\Delta x)^2 / \ell^2]$

→ Gaussian works.

$$T \sim S / (\ell^2 + (\Delta x)^2) -$$

→ Lorentzian fails

$$T \sim f(\Delta x) (\Delta x / \ell)^{-\alpha}$$

→ Power Law requires $\alpha > 3$.

Lesson: Tail of transition pdf can
have big effect on validity of
diffusive, random walk models.

⇒ 'Fat Tail' problem.

- N.B. Existence of second moment of transition probability enables application of Central Limit Thm:

(Simply Put) CLT:

As long as $\langle (\Delta x)^2 \rangle$ finite, then after N steps:

$$P_N(x) = \frac{\exp\left[-x^2 / N \langle \Delta x^2 \rangle\right]}{\left(N \langle \Delta x^2 \rangle\right)^{1/2}}$$

i.e. probability of location after N steps (1D) is Gaussian, with $\langle x^2 \rangle \sim N \langle \Delta x^2 \rangle$

- F.P.E. is conservative

i.e. 'particles' moved around, but not lost, up to boundary.

$$\partial n / \partial t = -\nabla \cdot \Gamma$$

$$\Gamma = -D \frac{\partial n}{\partial x} + Vn$$

$V=0$,
pure
diffusion

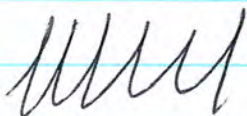
fundamentally, mathematical structure of random walk paths is rough

$$dx^2 \sim dt, \quad \Delta x \sim (\Delta t)^{1/2}$$

as compared to usual $\Delta x \sim \Delta t$

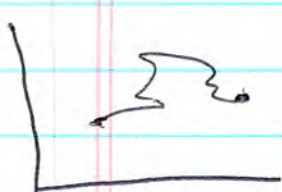
i.e. usual: $f(t+\Delta t) - f(t) \sim \Delta t$

diffn: $\begin{cases} f(t+\Delta t) - f(t) \sim \Delta t^{1/2} \\ f = \Delta x \end{cases} \rightarrow \text{not-differentiable}$

i.e. 

Ex: Brownian Motion

- classic example of diffusion arises in random walk of particle driven by thermal random kicks, and restricted by drag.



small particle:
 $l \rightarrow$ scale

$$\eta = \rho \nu = \rho v_{th} l_{mp}$$

viscosity

→ Diffusion has an \ominus H-Thm.

$$\frac{\partial n}{\partial t} = \frac{\partial}{\partial x} D \frac{\partial n}{\partial x}$$

closed system ($\Gamma \rightarrow 0$, on bndry)

$$\partial_t \int d^3x n^2 = -D \int d^3x (\partial n)^2$$

so $\partial_t \int d^3x n^2$ decreases unless

$\partial n = 0$, everywhere.

→ "S" = $-\int d^3x n^2$

$$m_p \frac{d\mathbf{v}}{dt} = -\nabla V + \tilde{\mathbf{F}}$$

\downarrow particle mass. \downarrow Stokes drag $\sim 6\pi\eta R$ \downarrow dim. \downarrow Brownian force random (additive) \rightarrow thermal noise \rightarrow thermal fluctuations.

n.b. Fluid exerts both drive (thermal fluctuations) and drag (β) on Brownian particle.

$\tilde{\mathbf{F}} \rightarrow$ Random Force / No Memory

$$\langle \tilde{\mathbf{F}}(t_1) \tilde{\mathbf{F}}(t_2) \rangle = \tilde{\mathbf{F}}^2 \tau_{\text{FC}} \delta(t_2 - t_1)$$

\downarrow
strength

$\tau_{\text{FC}} \rightarrow$ required for dimensions
 \rightarrow memory time of force necessarily shortest time in problem.

$$\langle \tilde{\mathbf{F}}^2 \rangle_{\omega} = \int e^{-i\omega(t_2 - t_1)} \langle \tilde{\mathbf{F}}(1) \tilde{\mathbf{F}}(2) \rangle d(t_2 - t_1)$$

$$= \tilde{\mathbf{F}}^2 \tau_{\text{FC}} \rightarrow \text{const} \rightarrow \text{"white noise"}$$

What is $\tilde{\mathbf{F}}^2$?

$$m_p \frac{dv}{dt} = -\beta v + \tilde{F}$$

steady state:

$$\beta \langle v^2 \rangle = \langle \tilde{F} \cdot v \rangle \quad \text{at } T.$$

Power
dissipated
by drag

Power input
by Brownian force.

but $m_p \langle v^2 \rangle \sim T \rightarrow$ both sets
thermal reservoir,
at T .

$$\langle v^2 \rangle \sim \frac{T}{m_p} \sim \frac{\langle \tilde{F} \cdot v \rangle}{\beta}$$

$$\text{but } v \sim \tilde{F}/\beta \quad (\text{SS})$$

$$\langle v^2 \rangle \sim \frac{T}{m_p} \sim \frac{\langle \tilde{F} \cdot v \rangle}{\beta} \sim \frac{\langle \tilde{F}^2 \rangle}{\beta^2}$$

\Rightarrow arrives at particularly simple form
of Fluctuation - Dissipation Theorem

ve

→ drag-induced energy dissipation balances fluctuation work at steady state, to maintain temperature T.

→ given any 2 of T, drag, fluctuations (force), can deduce third.

For diffusion of Brownian Particle:

$$D \sim \frac{\langle \Delta x^2 \rangle}{\Delta t} \sim \frac{v_{th}^2 (\Delta t)^2}{\Delta t} \sim v_{th}^2 \Delta t$$

$$\text{now } m_p \frac{dv}{dt} = -\beta v + \underline{F}$$

$$\frac{dv}{dt} = \frac{-\beta v + \underline{q}}{m_p}$$

velocity ~~remains~~ acts over time scale $\Delta t \sim m_p / \beta$ (~ skin colln time)

$$D \sim \langle v^2 \rangle \Delta t \sim \frac{T}{m_p} \frac{m_p}{\beta} \sim \frac{T}{\beta}$$

so,

$$D \sim \frac{T}{6\pi\eta r}$$

→ diffusivity, in space, of Brownian particle.

→ alternatively:

$$m_p \frac{dv}{dt} = -\beta v + \tilde{F}$$

$dv/dt = 0 \Rightarrow$ terminal velocity

$$\frac{dx}{dt} = \frac{\tilde{F}}{\beta}$$

$$dx \sim \int \frac{\tilde{F}}{\beta} dt$$

$$\langle dx^2 \rangle \sim \frac{\tilde{F}^2}{\beta^2} (\Delta t) t$$

$$\begin{aligned} \langle dx^2 \rangle &\sim \iint \frac{\tilde{F}^2}{\beta^2} dt dt' \\ &\sim \frac{\tilde{F}^2}{\beta^2} \Delta t t \end{aligned}$$

$$F = 0 \quad T: \quad \langle \tilde{F}^2 \rangle = \beta^2 T / m_p$$

$$\Delta t = m_p / \beta$$

$$\begin{aligned} \langle dx^2 \rangle &\sim \frac{\tilde{F}^2}{\beta^2} \Delta t t \sim \left(\frac{\beta T}{m_p} \frac{m_p}{\beta} \right) t \\ &\sim \beta T t \\ &\sim (T/\beta) t \quad \checkmark \end{aligned}$$

⇒ As usual, back to Basic Scales
in Random Walk, / Diffusion of
Brownian Particle

τ_c → self-correlation time of Brownian
Force

~ effectively → 0

White Noise ↔ band width → ∞

τ_c
 Δt → step time, velocity correlation
time

i.e., $\tilde{v} \Delta t \sim \tilde{v} \tau_c \sim v_{th} \tau_c \sim \Delta r$

$$\tau_c^{-1} \sim \beta / m_p$$

↑
spatial
step

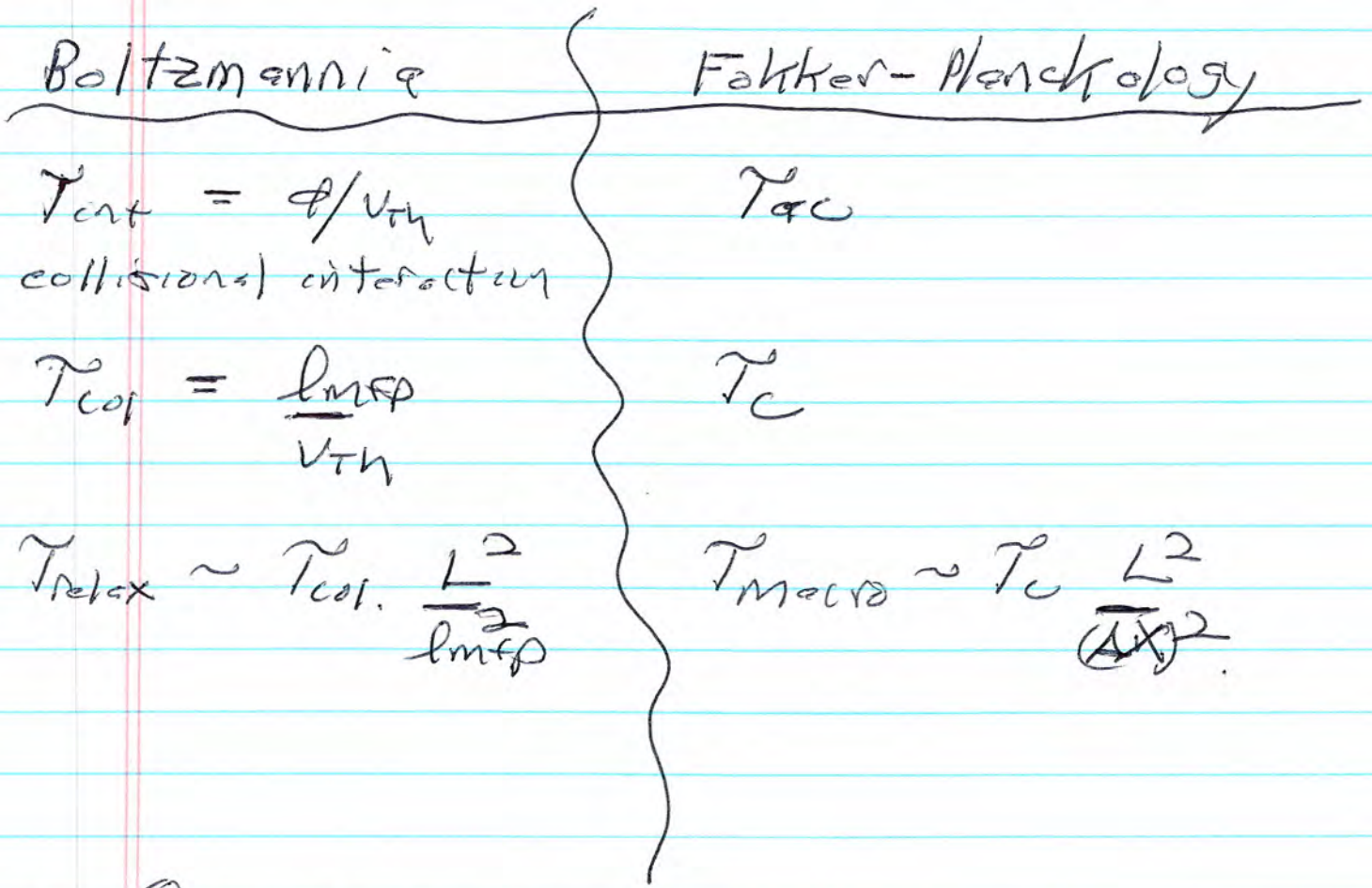
τ_d → macro-diffusion time

$$1/\tau_d \sim D/L^2 \sim T/\beta L^2$$

$$\tau_c / \tau_d \sim \frac{T}{\beta L^2} \frac{m_p}{\beta} \sim \frac{v_{th}^2}{L^2} \left(\frac{m_p}{\beta} \right)^2$$

$$\sim \frac{\Delta r^2}{L^2} \quad \downarrow$$

So - analogy:



N.B. ① Analogy, only!

② Note:

- Boltzmannie allows arbitrary collision

- F.P. $\underline{E} \Rightarrow$ weak glancing collision,
so $|\Delta p| \ll |\underline{p}|$

- interesting application: sedimentation



Brownian particles of size l , in fluid at T, ρ .

What is spatial distribution?

- particles random walk $\rightarrow T$
- " drift, due gravity

Profile: at st state \Rightarrow

$$n(z) = e^{-\frac{m g z}{T}}$$

Now,
$$m \rho \frac{d\mathbf{v}}{dt} = -\nabla V - m \rho g \hat{z} + \tilde{F}$$

or, in 1D:

$$\frac{d\mathbf{v}}{dt} = -\alpha \mathbf{v} - g + \tilde{a}$$

Now, at terminal velocity, drag and forces balance so i
(neglect transient)

$$\frac{dx}{dt} = -\frac{g}{\alpha} \hat{z} + \frac{v_{th}}{\alpha} \hat{z}$$

Consider \hat{z} direction, only, so:

$$\frac{dz}{dt} = -\frac{g}{\alpha} + \frac{v_{th}}{\alpha}$$

Now, F.P. E. \Rightarrow

$$\frac{\partial n}{\partial t} = -\frac{\partial}{\partial z} \left\{ \frac{\langle \Delta z \rangle}{\Delta t} n - D_z \frac{\partial n}{\partial z} \right\}$$

$$\left\langle \frac{\Delta z}{\Delta t} \right\rangle = \left\langle \frac{dz}{dt} \right\rangle = -\frac{g}{\alpha} = v \quad \rightarrow \text{drift speed}$$

$$D_z \sim \frac{\langle v^2 \Delta t \rangle}{\Delta t} \sim \frac{\langle \tilde{v} \rangle^2}{\alpha} \Delta t$$

\downarrow
fluctn. part

$$\sim \frac{\tilde{v}^2}{\alpha^2} \Delta t \sim \frac{\tilde{v}^2}{\alpha^3}$$

but:

$$D_z \sim \frac{v_{th}^2}{\alpha^3} \sim \frac{1}{m_p^2} \frac{f^2}{\omega^3} m_p^3 \sim T/B \quad \text{as used}$$

Now, $D = T/\beta$

$$V = -\frac{gmp}{\beta}$$

Observe: $\underline{V} = \frac{1}{\beta} \underline{F}$ $\underline{F} = -gmp \hat{z}$

\downarrow
force

$$= \mu \underline{F}$$

\downarrow
— mobility, here $= 1/\beta$

— generic to friction produced terminal free-fall.

$$D = T/\beta = \mu T \quad \checkmark$$

example of Einstein Relation

$$D = \mu T$$

relates D to mobility.
Mobility easy to calculate

generic to Brownian Motion type EOM!
and FDT

$$m_p \frac{d\underline{v}}{dt} = -\beta \underline{v} + \underline{f}$$

→ for steady state:

$$\frac{\partial n}{\partial t} = -\frac{\partial}{\partial z} \left\{ \frac{-mg}{\beta} n - \frac{T}{\beta} \frac{\partial n}{\partial z} \right\}$$

$$0 = \frac{-mg}{\beta} n - \frac{T}{\beta} \frac{\partial n}{\partial z}$$

Balance is:
 - vertical drift
 - $\frac{us}{\beta}$
 - upper diffusion

$$\Rightarrow n = \exp\left[-\frac{mgz}{T}\right] \checkmark$$

Of course can obtain time relaxation to equilibrium, as well.

→



Full evolution:
 → vertical sedimentation
 → radial diffusion at D_n .

Key PT: - drift, diffusion relation

- stationarity from zero flux condition.

↳ Multiplicative Noise

Additive: $\frac{dV}{dt} = -\alpha V + \tilde{\eta}(t)$

Multiplicative: $\frac{dn}{dt} = \gamma n - \alpha n^2$

$\gamma = \gamma_0 + \tilde{\gamma}$

i.e.

- logistic population equation
- noise enters as multiplier on growth rate.

$\langle \tilde{\gamma}(t) \tilde{\gamma}(t') \rangle = \gamma_0^2 \gamma_{rel} \delta(t-t')$

- (randm variable) $n \rightarrow$ unusual behavior

Logistic System

$\frac{dn}{dt} = \gamma n - \alpha n^2$

↓
Malthusian growth

↳ { saturation
by competition

2 equilibria: $n=0$ unstable
 $n = \gamma/\alpha$ stable

Now, $\gamma = \gamma_0 + \tilde{\gamma}(t)$
 $\alpha = 1$
↳ i.e. variability in conditions, food supply etc.

18, need distribution of populations

⇒ Fokker-Planck Eqn. for $f(n, t)$!

can immediately write

$$\frac{dn}{dt} = (\gamma_0 + \underbrace{\tilde{\gamma}}_{\text{random}})n - \alpha n^2$$

$$\frac{\partial f}{\partial t} = -\frac{\partial}{\partial n} \left\{ \left\langle \frac{dn}{dt} \right\rangle f - \frac{\partial}{\partial n} D f \right\}$$

$$D = \frac{\langle \delta n^2 \rangle}{2\Delta t}$$

Now, $\langle dn/dt \rangle = \gamma_0 n - \alpha n^2$

$$\frac{d \tilde{\gamma} n}{dt} = \tilde{\gamma} n$$

$$\langle \tilde{\gamma}(t_1) \tilde{\gamma}(t_2) \rangle = \tilde{\gamma}^2 \gamma_{00} \delta(t_2 - t_1)$$

$$\Rightarrow D = \frac{\langle \delta n \delta n \rangle}{2\Delta t}$$

$$= \int dt_1 \int dt_2 \frac{\tilde{\gamma}^2 \gamma_{00} (t_2 - t_1)}{2} n^2 / \Delta t$$

$$\approx n^2 \sigma^2$$

$$\begin{aligned} \overline{\sigma^2} &\equiv \int \langle \delta(t_1) \delta(t_2) \rangle dt \\ &\sim \overline{\sigma^2} T_{el} \\ &\sim 1/T \quad \checkmark \end{aligned}$$

but now D nonlinear!! \rightarrow consequence
 \Rightarrow multiplicative character

$$\frac{\partial f}{\partial t} = \frac{\partial}{\partial n} \left\{ (\gamma_0 n - n^2) f - \frac{\partial}{\partial n} (n^2 \sigma^2 f) \right\}$$

\Rightarrow Fokker-Planck Eqn. for f .

Stationary state \Rightarrow

$$(\gamma_0 n - n^2) f - \frac{\partial}{\partial n} (n^2 \sigma^2 f) = 0$$

$$A(n) f - \frac{\partial}{\partial n} (B f) = 0$$

\Rightarrow

$$f \sim \frac{1}{B(n)} \exp \left[\int dn' \frac{A(n')}{B(n')} \right]$$

Now,

$$\int \delta n' \left[\frac{\delta_0 n' - n'^2}{\sigma^2 n'^2} \right] = \int \delta n' \left[\frac{\delta_0}{\sigma^2 n'} - \frac{1}{\sigma^2} \right]$$

$$= \frac{\delta_0}{\sigma^2} - \frac{1}{\sigma^2} + \frac{\delta_0}{\sigma^2} \ln(n)$$

$$\Rightarrow \boxed{F = \frac{1}{\sigma^2 n^2} n^{\delta_0/\sigma^2} \exp\left[-n/\sigma^2\right]}$$

and need:

PDF

$$\frac{\delta_0}{\sigma^2} - 2 > -1$$

to assure
integrability

\Rightarrow

$$\boxed{\delta_0/\sigma^2 > 1}$$

fluctuations in
growth cannot
be too large.

Meaning:

plot $f(n)$



$\frac{\delta_0}{\sigma} = 1$ is determ.
or b.r.n.