

Trick to calculate: start with diffusion equation

$$x^2 \frac{\partial P}{\partial t} = \frac{\partial}{\partial t} x^2 P = D x^2 \frac{\partial^2 P}{\partial x^2} \quad \Big| \quad \int_{-\infty}^{\infty} dx.$$

$$\frac{d}{dt} \langle x^2 \rangle = D \left\{ \left[x^2 P' \right]_{-\infty}^{\infty} - \int_{-\infty}^{\infty} 2x \frac{\partial P}{\partial x} dx \right\} = D \left\{ - \left[2xP \right]_{-\infty}^{\infty} + \int_{-\infty}^{\infty} 2P dx \right\} = 2D$$

$$\Rightarrow \langle x^2 \rangle = 2Dt.$$

Fick's derivation of the diffusion equation:

(1) Continuity equation: Given a probability density $P(\underline{r}, t)$ with

$$\int_{-\infty}^{\infty} P(\underline{r}, t) dV = 1, \text{ and the probability flux } \underline{S}(\underline{r}, t):$$

$$\oint_{\partial V} \underline{S}(\underline{r}, t) d\underline{A} = - \frac{d}{dt} \iiint_V P(\underline{r}, t) dV = - \frac{d}{dt} \psi(t)$$

where $\psi(t)$ is the survival probability.

With the divergence theorem (Gauss' theorem):

$$\oint_{\partial V} \underline{S}(\underline{r}, t) d\underline{A} = \iiint_V \nabla \cdot \underline{S}(\underline{r}, t) dV$$

we obtain the integral form of the continuity equation:

$$\iiint_V \nabla \cdot \underline{S}(\underline{r}, t) dV = - \frac{d}{dt} \iiint_V P(\underline{r}, t) dV = - \iiint_V \frac{\partial P(\underline{r}, t)}{\partial t} dV$$

This relation is valid $\forall (V, \partial V) \Rightarrow$ differential form:

$$\underline{\underline{\frac{\partial}{\partial t} P(\underline{r}, t) = - \nabla \cdot \underline{S}(\underline{r}, t) = - \operatorname{div} \underline{S}(\underline{r}, t). \quad \text{continuity equation}}}$$

(2) Fick's first law:

$$\underline{\underline{\underline{S} = - D \nabla P(\underline{r}, t)}}$$

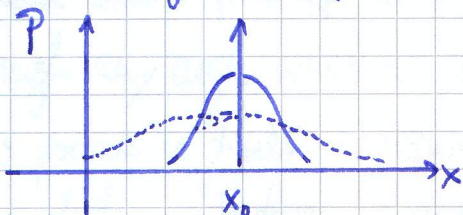
the flux is proportional to the gradient of the probability density function P .

$$\Rightarrow \frac{\partial}{\partial t} P(x, t) = -\nabla \cdot \underline{S}(x, t) = D \nabla^2 P(x, t) \quad \text{Fick's second law}$$

In the following, for simplicity we will deal with the one-dimensional case.

Boundary value problems:

(1) Reflecting boundary at $x=0$ & initial condition $P_0(x) = \delta(x-x_0)$, $x_0 > 0$



$$\left. \frac{\partial P(x, t)}{\partial x} \right|_{x=0} = 0$$

Solution: (a) Standard method: Laplace transform & solution of ODE in x .

(b) Method of images: The portion of the probability "leaking" out to $x < 0$ of the Gaussian $(4\pi Dt)^{-1/2} \exp(-[x-x_0]^2/(4Dt))$ is compensated by the influx of probability from a ~~mirror~~ source at $-x_0$. The resulting Green's function solves the boundary value problem:

$$Q(x, t) = G(x-x_0, t) + G(x+x_0, t) \quad \text{where } G(x, t) = \frac{1}{\sqrt{4\pi Dt}} e^{-x^2/4Dt}$$

Moreover Q is normalised. Thus the image Q is the sought-after solution.

Diffusion on average moves the particle away from the origin:

$$\langle x \rangle = \int_0^\infty Q(x, t) dx = 2 \sqrt{\frac{Dt}{\pi}} e^{-x_0^2/4Dt} + x_0 \operatorname{erf}\left(\frac{x_0}{\sqrt{4Dt}}\right) \sim \sqrt{\frac{4Dt}{\pi}}$$

(2) Absorbing boundary conditions at $x=0$:

$$P(0, t) = 0$$

Image solution $Q(x, t) = G(x-x_0, t) - G(x+x_0, t)$.

$$\text{Survival probability: } \mathcal{V}(t) = \int_0^\infty Q(x, t) dx = \operatorname{erf}\left(\frac{x_0}{\sqrt{4Dt}}\right) \sim \frac{x_0}{\sqrt{\pi Dt}}$$

III. RANDOM WALKS.

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1905 Karl Pearson poses the problem of the random walk, being interested in the dynamics how mosquito populations invade cleared jungle regions:

"Can any of your readers refer me to a work wherein I should find the solution to the following problem [...]:

A man starts from a point O and walks l yards in a straight line; he then turns through any angle whatever and walks l yards in a second straight line. He repeats this process n times. I require the probability that after these n stretches he is at a distance between r and $r+dr$ from his starting point, O ."

Lord Rayleigh pointed out one week later that in the limit of large n the solution is a Gaussian.

Considers a jump process on a one-dimensional lattice with spacing a . At each step the probability to jump left or right is equal.

$\leadsto$ Probability to reach site m after N steps ($N \pm m$ are always even.)

$$g(m, N) = \left(\frac{1}{2}\right)^N \frac{N!}{\underbrace{\left[\frac{1}{2}(N-m)\right]!}_{\frac{N-m}{2} \text{ jumps to left}} \underbrace{\left[\frac{1}{2}(N+m)\right]!}_{\frac{N+m}{2} \text{ jumps to right}}}$$

Stirling's formula: $\log N! = \left(N + \frac{1}{2}\right) \log N - N + \frac{1}{2} \log 2\pi + O(N^{-1})$

Expansion of log: $\log(1+z) \sim z - \frac{z^2}{2}, z \ll 1$.

$$\Rightarrow \log g(m, N) \approx -\frac{1}{2} \log N + \log 2 - \frac{1}{2} \log 2\pi - \frac{m^2}{2N}$$

$$g(m, N) \approx \sqrt{\frac{2}{\pi N}} \exp\left(-\frac{m^2}{2N}\right)$$

m is either always even (N even) or always odd (N odd)

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$\Rightarrow$ probability to meet the walker at site m :

$$\underline{\underline{P(m, N) = \frac{1 + (-1)^{N-m}}{2} g(m, N) \approx \frac{1 + (-1)^{N-m}}{\sqrt{2\pi N}} \exp\left(-\frac{m^2}{2N}\right). \quad (*)}}$$

Note that $P(m, N) = 0$ for $|m| > N$ for exact expression, but not for approximation. This result is one form of the central limit theorem (CLT): the (normalised) sum of independent random variables with finite variance is well approximated by a random variable with a Gaussian distribution in the limit of large numbers.

Even for $N=10$ the approximation (*) containing the Gaussian is almost exact (normalisation is 0.999594).

The factor $\frac{1 + (-1)^{N-m}}{2}$ is essential for the normalisation.

Continuum limit:

Position of the walker is $x = ma$:

$$P(x, N) \approx \sqrt{\frac{1}{2\pi N a^2}} \exp\left(-\frac{x^2}{2Na^2}\right)$$

↑
from Jacobien

Time $t = N\Delta t$ and diffusion constant $K = \frac{a^2}{2\Delta t}$

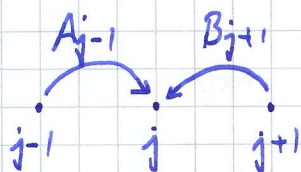
$$\underline{\underline{\Rightarrow P(x, t) \approx \sqrt{\frac{1}{4\pi K t}} \exp\left(-\frac{x^2}{4Kt}\right).}}$$

IV. THE FOKKER-PLANCK EQUATION.

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Let us start from a random walk, but now the probability to jump left or right depends on the walker's position: we start from the discrete master equation

$$W_j(t+\Delta t) = A_{j-1} W_{j-1}(t) + B_{j+1} W_{j+1}(t) \quad \text{Pr to find walker at } j \text{ at } t+\Delta t$$



$$A_j + B_j = 1 \text{ jump/w } \text{Pr} = 1.$$

Taylor expansions: $W_j(t+\Delta t) \approx W_j(t) + \Delta t \frac{\partial}{\partial t} W_j(t)$

$$A_{j-1} W_{j-1} \approx A(x) W(x) - \Delta x \frac{\partial}{\partial x} A(x) W(x) + \frac{(\Delta x)^2}{2} \frac{\partial^2}{\partial x^2} A(x) W(x)$$

$$B_{j+1} W_{j+1} \approx B(x) W(x) + \Delta x \frac{\partial}{\partial x} B(x) W(x) + \frac{(\Delta x)^2}{2} \frac{\partial^2}{\partial x^2} B(x) W(x)$$

$$\Rightarrow W(x, t) + \Delta t \frac{\partial W}{\partial t} = \underbrace{[A(x) + B(x)]}_{=1} W(x, t) + \Delta x \frac{\partial}{\partial x} (B(x) - A(x)) W(x, t) + \frac{(\Delta x)^2}{2} \frac{\partial^2}{\partial x^2} [A(x) + B(x)] W(x, t)$$

$$\frac{\partial}{\partial t} W(x, t) = \frac{\Delta x}{\Delta t} \frac{\partial}{\partial x} [B(x) - A(x)] W(x, t) + \frac{(\Delta x)^2}{2 \Delta t} \frac{\partial^2}{\partial x^2} W(x, t)$$

Continuum limit $\Delta x \rightarrow 0, \Delta t \rightarrow 0$:

$$\frac{\partial}{\partial t} P(x, t) = \frac{\partial}{\partial x} \frac{V'(x)}{m\eta} P(x, t) + K \frac{\partial^2}{\partial x^2} P(x, t) \quad \text{Fokker-Planck equation}$$

$$\left. \begin{aligned} \lim_{\Delta x \rightarrow 0, \Delta t \rightarrow 0} \frac{\Delta x}{\Delta t} [B(x) - A(x)] &= \frac{V'(x)}{m\eta} \\ \lim_{\Delta x \rightarrow 0, \Delta t \rightarrow 0} \frac{(\Delta x)^2}{2 \Delta t} &= K \end{aligned} \right\} \text{for proper limit we must have that } B(x) - A(x) \approx \Delta x$$

$F(x) = -V'(x)$ is the external force. η is the friction coefficient.

In a confining potential the long-time limit of $P(x, t)$ reaches the equilibrium distribution:

$$P_{eq}(x) = \mathcal{N} \exp\left(-\frac{V(x)}{\ln BT}\right).$$

$$\frac{\partial}{\partial t} P(x, t) = 0 \Rightarrow 0 = \left(\frac{\partial}{\partial x} \frac{V'(x)}{m\eta} + K \frac{\partial^2}{\partial x^2} \right) P(x, t)$$

$$\leadsto 0 = \left(\frac{V'(x)}{m\eta} + K \frac{\partial}{\partial x} \right) P_{eq}(x)$$

$$\rightarrow P_{eq}(x) = \mathcal{N} \exp\left(-\frac{V'(x)}{K m \eta}\right)$$

$$\Rightarrow \underline{\underline{K = \frac{\ln BT}{m\eta}}} \quad \text{Einstein - Stokes - Smolukowski relation} \\ \text{(fluctuation - dissipation relation).}$$

Constant external force F_0 :

$$\frac{\partial}{\partial t} P(x, t) = -\frac{F_0}{m\eta} \frac{\partial}{\partial x} P(x, t) + K \frac{\partial^2}{\partial x^2} P(x, t) \quad \text{sometimes called diffusion-advection eq.}$$

(i.) Similarity solution with Galilei (wave) variable $\xi = x - \frac{F_0}{m\eta} t$

$$P(x, t) = G\left(x - \frac{F_0}{m\eta} t, t\right) \therefore G(x, t) = \frac{1}{\sqrt{4\pi K t}} \exp\left(-\frac{x^2}{4K t}\right).$$

(ii.) Second Einstein relation (linear response):

$$\frac{d}{dt} \langle x(t) \rangle = \frac{F_0}{m\eta} \Rightarrow \langle x(t) \rangle = \frac{F_0}{m\eta} t = \frac{F_0 K}{\ln BT} t$$

w/o force we know that: $\langle x^2(t) \rangle_0 = 2Kt$

$$\Rightarrow \underline{\underline{\langle x(t) \rangle_{F_0} = \frac{F_0}{2 \ln BT} \langle x^2(t) \rangle_0.}}$$

V. The continuous time random walk process

EW Montroll, GH Weiss, J Math Phys 6, 167 (1965).

H Scher, M Lax, Phys Rev B 7, 4502 (1973).

MF Shlesinger, J Stat Phys 10, 421 (1974).

H Scher, EW Montroll, Phys Rev B 12, 2455 (1975).

J Klafter, R Silbey, Phys Rev Lett 44, 55 (1980).

Basic ingredient: Waiting time distribution $\varphi(t)$

After each jump the walker waits for a random time drawn from the probability density function $\varphi(t)$. We use $\varphi(u) = \langle e^{-ut} \rangle$

Sticking probability for not moving:

$$\phi(t) = \int_t^\infty \varphi(t') dt' = 1 - \int_0^t \varphi(t') dt' \rightarrow \phi(u) = \frac{1}{u} - \frac{\varphi(u)}{u} = \frac{1 - \varphi(u)}{u}$$

PDF of occurrence of i^{th} step at time $t = t_1 + t_2 + \dots + t_i$

$$\begin{aligned} \mathcal{L}\{\varphi_i(t)\} &= \underline{\varphi_i(u)} = \langle e^{-ut} \rangle = \langle e^{-u(t_1 + t_2 + \dots + t_i)} \rangle \stackrel{\text{independent}}{\underset{t_i}{=}} \\ &= \langle e^{-ut_1} \rangle \langle e^{-ut_2} \rangle \dots \langle e^{-ut_i} \rangle = \langle e^{-ut} \rangle^i = \underline{\varphi^i(u)} \end{aligned}$$

$\Rightarrow$ Probability that the walker has jumped exactly i times up to time t :

$$Q_i(t) = \int_0^t \varphi_i(t') \phi(t-t') dt'$$

$$\rightarrow Q_i(u) = \varphi_i(u) \phi(u) = \frac{1 - \varphi(u)}{u} \varphi^i(u).$$

$\Rightarrow$ Probability to find walker at site n at time t :

$$\mathcal{P}(n, t) = \sum_{i=0}^{\infty} p_i(n) Q_i(t) \therefore p_i(n) \text{ is probability at site } n \text{ after } i^{\text{th}} \text{ step.}$$

$$\leadsto \mathcal{P}(n, u) = \frac{1 - \varphi(u)}{u} \sum_{i=0}^{\infty} p_i(n) \varphi^i(u) = \frac{1 - \varphi(u)}{u} p_0(n) + \frac{1 - \varphi(u)}{u} \sum_{i=1}^{\infty} p_i(n) \varphi^i(u)$$

(**)

Continuum limit:

$p_0(x) = \delta(x)$ initial condition

$p_{i+1}(x) = p_i(x) + \frac{a^2}{2} \frac{\partial^2}{\partial x^2} p_i(x) + \dots$ according to our results from chapter IV.

$$\Rightarrow \mathcal{P}(x, u) = \frac{1-\varphi(u)}{u} \delta(x) + \frac{1-\varphi(u)}{u} \sum_{i=1}^{\infty} \left(p_{i-1}(x) + \frac{a^2}{2} \frac{\partial^2}{\partial x^2} p_{i-1}(x) \right) \varphi^i(u)$$

$$\text{but from (**) } \frac{1-\varphi(u)}{u} \sum_{i=1}^{\infty} p_{i-1}(x) \varphi^i(u) = \frac{1-\varphi(u)}{u} \sum_{i=0}^{\infty} p_i(x) \varphi^{i+1}(u) \\ = \mathcal{P}(x, u) \varphi(u)$$

$$\Rightarrow \mathcal{P}(x, u) = \frac{1-\varphi(u)}{u} \delta(x) + \varphi(u) \left(\mathcal{P}(x, u) + \frac{a^2}{2} \frac{\partial^2}{\partial x^2} \mathcal{P}(x, u) \right)$$

$$\underline{\underline{\mathcal{P}(x, u) - \frac{\delta(x)}{u} = \frac{\varphi(u)}{1-\varphi(u)} \frac{a^2}{2} \frac{\partial^2}{\partial x^2} \mathcal{P}(x, u)}} \quad \text{CTRW with arbitrary } \varphi(x) \text{ is the continuous space approximation.}$$

(1.) Back to the diffusion equation:

Sharp waiting time PDF: $\varphi(t) = \delta(t - \tau) \Rightarrow \varphi(u) = e^{-u\tau} \approx 1 - u\tau + \dots$

many step $\equiv$ long time corresponds to short $u \sim$ powers $\sim (u\tau)^2 \dots$ neglected

Same result for Poissonian $\varphi(t)$:

$$\varphi(t) = \tau^{-1} e^{-t/\tau} \Rightarrow \varphi(u) = \frac{1}{u + \tau^{-1}} \approx 1 - u\tau + \dots$$

In fact, this behaviour is universal, as long as the characteristic waiting time ~~time~~ remains finite:

$$\langle \tau \rangle = \int_0^{\infty} \varphi(t) t = \int_0^{\infty} t \varphi(t) e^{-ut} dt \Big|_{u=0} = - \frac{d\varphi(u)}{du} \Big|_{u=0}$$

$\Rightarrow$ for finite $\langle \tau \rangle$ we have $\varphi(u) \approx 1 - \langle \tau \rangle u + \dots$

Up to first order in τ : $\frac{\varphi(u)}{1-\varphi(u)} = \frac{1-u\tau}{u\tau} = \frac{1}{u\tau} - 1$

$$\Rightarrow \mathcal{P}(x, u) - \frac{\delta(x)}{u} = \left(\frac{1}{u\tau} - 1\right) \frac{a^2}{2} \frac{\partial^2}{\partial x^2} \mathcal{P}(x, u)$$

The diffusion constant is $K = \lim_{a \rightarrow 0, \tau \rightarrow 0} \frac{a^2}{2\tau}$

$\Rightarrow$ the term $\lim_{a \rightarrow 0, \tau \rightarrow 0} a^2 \rightarrow 0$

$$\Rightarrow \mathcal{P}(x, u) - \frac{\delta(x)}{u} = \frac{1}{u} K \frac{\partial^2}{\partial x^2} \mathcal{P}(x, u)$$

After inverse Laplace transformation:

$$\mathcal{P}(x, t) - \delta(x) = \int_0^t K \frac{\partial^2}{\partial x^2} \mathcal{P}(x, t') dt' \quad \text{integral form of diffusion eq.}$$

$$\Rightarrow \frac{\partial}{\partial t} \mathcal{P}(x, t) = K \frac{\partial^2}{\partial x^2} \mathcal{P}(x, t) \quad \checkmark$$

(2) Diverging characteristic waiting time:

Assume the form $\varphi(u) = e^{-(u\tau)^\alpha}$, $0 < \alpha < 1$ for the waiting time PDF. This is actually the characteristic function of a one-sided Lévy stable law, resulting from a generalisation of the CLT for random variables with diverging variance. The leading order of the Laplace inversion is $\varphi(t) \sim t^\alpha / t^{1+\alpha}$.

In particular, we see that $\langle \tau \rangle = \int_0^\infty t \varphi(t) dt = - \frac{d\varphi(u)}{du} \Big|_{u=0} = \tau^\alpha u^{\alpha-1} e^{-(u\tau)^\alpha} \Big|_0 \rightarrow \infty$.

This process is scale-free, i.e. there is no time scale separating single jump events from the limit of a large number of jumps $\Rightarrow$ ageing, non-ergodicity.

In Laplace space, we find

$$\mathcal{P}(x, u) - \frac{\delta(x)}{u} = \frac{1}{u^\alpha} K_\alpha \frac{\partial^2}{\partial x^2} \mathcal{P}(x, t) \therefore K_\alpha = \lim_{a \rightarrow 0, \tau \rightarrow 0} \frac{a^2}{2\tau^\alpha} \therefore [K_\alpha] = \frac{cm^2}{\text{sec}^\alpha} \cdot (*)$$

Mean squared displacement: Integrate over $\int x^2 dx$:

$$\langle x^2(u) \rangle = \frac{2K_\alpha}{u} \Rightarrow \underline{\underline{\langle x^2(t) \rangle = \frac{2K_\alpha}{\Gamma(1+\alpha)} t^\alpha}}$$

Processes with $\langle x^2 \rangle \sim t^\alpha$, $\alpha \neq 0$ are called anomalous diffusion.

$0 < \alpha < 1$: subdiffusion

There also exists superdiffusion with $\alpha > 1$ ($\alpha=2$: ballistic, wave-like)

Relaxation of single Fourier modes:

Fourier transform of eq (*):

$$\mathcal{P}(k, u) - \frac{1}{u} = -K_\alpha \frac{k^2}{u^\alpha} \mathcal{P}(k, u)$$

$$\sim \mathcal{P}(k, u) = \frac{1}{u + K_\alpha k^2 u^{1-\alpha}} = \frac{1}{u + u^{1-\alpha} \tau_*^{-\alpha}} \equiv \phi(u) \quad \tau_*^{-\alpha} \equiv K_\alpha k^2 \text{ is a time scale @ } k \text{ fixed.}$$

This Laplace image defines the Mittag-Leffler function:

$$\phi(t) = E_\alpha(-[t/\tau_*]^\alpha) = \sum_{n=0}^{\infty} \frac{(-[t/\tau_*]^\alpha)^n}{\Gamma(1+\alpha n)} \sim 1 - \frac{(t/\tau_*)^\alpha}{\Gamma(1+\alpha)}$$

In the limit $\alpha=1$ we recover the standard exponential function

For $0 < \alpha < 1$ the Mittag-Leffler function has the asymptotic expansion:

$$\phi(t) = \sum_{n=1}^{\infty} \frac{(-1)^{n-1} (\tau_*/t)^{\alpha n}}{\Gamma(1-\alpha n)} \sim \frac{(\tau_*/t)^\alpha}{\Gamma(1-\alpha)}$$

The Mittag-Leffler function interpolates between an initial stretched exponential law $\phi \sim \exp(-\frac{(t/\tau_*)^\alpha}{\Gamma(1+\alpha)})$ and a terminal inverse power-law.

For $\alpha=1/2$, the Mittag-Leffler function reduces to:

$$E_{1/2}(-[t/\tau_*]^{1/2}) = e^{t/\tau_*} \operatorname{erfc}\left(\left[\frac{t}{\tau_*}\right]^{1/2}\right).$$

An alternative derivation of CTRW with a jump length distribution $\lambda(x)$. 37

Jump length distribution $\lambda(x)$ can have finite or infinite variance

$$\langle \delta x^2 \rangle = \int_{-\infty}^{\infty} x^2 \lambda(x) dx.$$

Considers the PDF of just having arrived at x at time t , $\eta(x, t)$. This PDF fulfills the generalised master equation:

$$\eta(x, t) = \int_{-\infty}^{\infty} dx' \int_0^t dt' \eta(x', t') \lambda(x-x') \varphi(t-t') + \underbrace{\delta(x) \delta(t)}_{\text{initial condition needed @ } t=0}$$

$\eta(x, t)$ is related to the position PDF $P(x, t)$ via

$$P(x, t) = \int_0^t \eta(x, t') \varphi(t-t') dt' \text{ of having arrived at } t' \text{ and not having moved since.}$$

$$\Rightarrow P(k, u) = \eta(k, u) \varphi(u) = \frac{1 - \varphi(u)}{u} \left\{ \frac{1}{1 - \lambda(k) \varphi(u)} \right\}$$

$$\underline{P(k, u) = \frac{1 - \varphi(u)}{u} \frac{1}{1 - \lambda(k) \varphi(u)}}$$

(1.) Gaussian jump length distribution:

$$\lambda(k) = \exp\left(-\frac{1}{2} \sigma^2 k^2\right) \sim 1 - \frac{1}{2} \sigma^2 k^2; \text{ with } \varphi(u) \sim 1 - u\tau$$

$$\Rightarrow 1 - \lambda(k) \varphi(u) \sim 1 - \left(1 - \frac{1}{2} \sigma^2 k^2\right)(1 - u\tau) \sim u\tau + \frac{\sigma^2}{2} k^2$$

$$\Rightarrow P(k, u) = \frac{u}{u\tau + \frac{\sigma^2}{2} k^2} = \frac{1}{u + \frac{\sigma^2}{2\tau} k^2} \therefore K = \lim_{\sigma \rightarrow 0, \tau \rightarrow 0} \frac{\sigma^2}{2\tau}$$

$$\sim P(x, t) = \frac{1}{\sqrt{4\pi K t}} \exp\left(-\frac{x^2}{4K t}\right).$$

(2.) α -stable jump length distribution:

$$\lambda(k) = \exp\left(-\frac{\sigma^\mu}{\mu} |k|^\mu\right) \sim 1 - \frac{\sigma^\mu}{\mu} |k|^\mu \leadsto \langle \delta x^2 \rangle \rightarrow \infty, \lambda(x) \sim \frac{\sigma^\mu}{|x|^{1+\mu}}$$

$$P(k, u) = \frac{1}{u + K^\mu |k|^\mu} \leadsto P(k, t) = \exp(-K^\mu |k|^\mu t)$$

$$\leadsto P(x, t) \sim \frac{K^\mu t}{|x|^{1+\mu}} \quad \text{"Lévy flight."}$$

Question: What is the dynamic equation for $P(x, t)$ when in its Fourier-Laplace transform terms such as u^α and $|k|^\mu$ occur?

$\leadsto$ Fractional calculus.

Fractional order differentiation:

1695 in a letter to GA de l'Hospital, GW Leibniz writes: Thus it follows that $d^{1/2}x$ will be equal to $x \sqrt{dx:x}$, an apparent paradox, from which one day useful consequences will be drawn."

(1.) Phenomenological introduction of fractional order differentiation:

$$\frac{d^m x^n}{dx^n} = \frac{n!}{(n-m)!} x^{n-m} \quad \text{for integers } m, n$$

generalisation through introduction of Γ function:

$$\frac{d^\mu x^\nu}{dx^\nu} \equiv \frac{\Gamma(\nu+1)}{\Gamma(\nu-\mu+1)} x^{\nu-\mu} \quad \text{for real-valued } \mu, \nu$$

NB: In this definition the derivative of a constant is not zero:

$$\frac{d^\mu x^0}{dx^\nu} = \frac{1}{\Gamma(1-\mu)} x^{-\mu}$$

Application to exponential function:

$$\frac{d^\mu e^{-t}}{dt^\mu} = \frac{d^\mu}{dt^\mu} \sum_{n=0}^{\infty} \frac{(-t)^n}{\Gamma(1+n)} = \sum_{n=0}^{\infty} \frac{(-1)^n t^{n-\mu}}{\Gamma(n-\mu+1)} = t^{-\mu} e^{-t} \gamma^*(-\mu, -t)$$

incomplete γ function

As $\gamma^*(-n, y) = y^n \leadsto$ formula reduces to well-known differentiation of exp.