

Trick to calculate: start with diffusion equation

$$x^2 \frac{\partial p}{\partial x} = \frac{\partial}{\partial x} x^2 p = \frac{\partial}{\partial x} \left( \int_{-\infty}^{\infty} x^2 p^2 dx \right) \quad \Bigg| \quad \int_{-\infty}^{\infty} dx$$

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

Fick's derivation of the diffusion equation:

(1) Continuity equation: Given a probability density  $P(x, t)$  with  $\int_{-\infty}^{\infty} P(x, t) dx = 1$ , and the probability flux  $\bar{S}(x, t)$ :

$$\oint_{\partial V} \bar{S}(x, t) d\bar{A} = - \frac{d}{dt} \int_V P(x, t) dV = - \frac{d}{dt} \langle x \rangle$$

where  $\langle x \rangle$  is the survival probability. With the divergence theorem (Gauss' theorem):

$$\oint_{\partial V} \bar{S}(x, t) d\bar{A} = \int_V \nabla \cdot \bar{S}(x, t) dV$$

We obtain the integral form of the continuity equation:

$$\int_V \nabla \cdot \bar{S}(x, t) dV = - \frac{d}{dt} \int_V P(x, t) dV = - \frac{d}{dt} \int_V \frac{\partial P(x, t)}{\partial t} dV$$

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

$$\frac{\partial}{\partial t} P(x, t) = - \nabla \cdot \bar{S}(x, t) = - \text{div } \bar{S}(x, t) \quad \text{continuity equation}$$

(2) Fick's first law:

$$\bar{S} = - \mathcal{D} \nabla P(x, t)$$

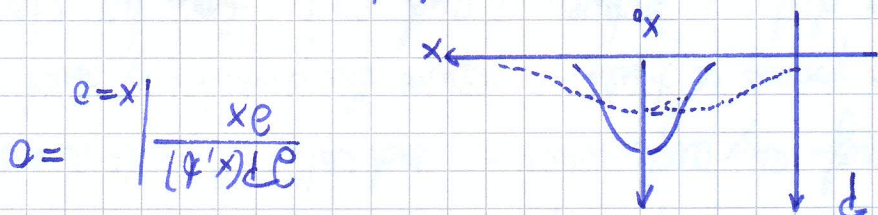
the flux is proportional to the gradient of the probability density function  $P$ .

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

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

### Boundary value problems:

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



Solution: (a) Standard method: Laplace transform & solution of  $\partial_t E$  in  $x$ .

(b) Method of images: The portion of the probability "leaking" out to  $x < 0$  of the Gaussian  $(\frac{1}{\sqrt{4\pi Dt}})^{-1/2} \exp(-[x-x_0]^2/(4Dt))$  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 normalized. Thus the image  $Q$  is the sought-after solution.

Diffusion on average moves the particle away from the origin:  

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

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

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

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

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

III. RANDOM WALKS.

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  $\theta$  and walks  $\lambda$  yards in a straight line; he then turns through any angle whatever and walks  $\lambda$  yards in a second straight line. He repeats this process  $n$  times. I require the probability that after these  $n$  strides he is at a distance between  $r$  and  $r+dr$  from his starting point,  $\theta$ .

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

$$g(m, N) = \binom{N}{\frac{N+m}{2}} \left(\frac{1}{2}\right)^N \frac{N!}{\left[\frac{1}{2}(N-m)\right]! \left[\frac{1}{2}(N+m)\right]!}$$

$\underbrace{\hspace{10em}}_{N-m \text{ jumps to left}} \quad \underbrace{\hspace{10em}}_{\frac{N+m}{2} \text{ jumps to right}}$

Stirling's formula:  $\log N! = (N + \frac{1}{2}) \log N - N + \frac{1}{2} \log 2\pi + \theta(N^{-1})$   
 Expansion of  $\log$ :  $\log(1+z) \approx z - \frac{z^2}{2} + \dots, z \ll 1$ .

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

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

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



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

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

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.

Continuous 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 Jacobian

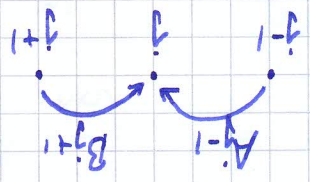
Time  $t = Na^2$  and diffusion constant  $K = \frac{a^2}{2\Delta t}$

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

IV. THE FOKKER-PLANCK EQUATION.

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

For to find walker at  $j$  at  $t+\Delta t$   
 $A_j + B_j = 1$  jump in  $\mathbb{R} = 1$ .



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

$$\frac{\partial}{\partial t} P(x,t) = \frac{\partial}{\partial x} \left[ \frac{V(x)}{m\eta} P(x,t) \right] + K \frac{\partial^2}{\partial x^2} P(x,t)$$

Fokker-Planck equation

$$\lim_{\Delta x \rightarrow 0, \Delta t \rightarrow 0} \frac{\Delta x}{\Delta t} [B(x)-A(x)] = \frac{V(x)}{m\eta}$$

$$\lim_{\Delta x \rightarrow 0, \Delta t \rightarrow 0} \frac{(\Delta x)^2}{2 \Delta t} = K$$

for proper limit we must have that  $B(x) - A(x) \approx \Delta x$

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

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

distribution:

$$P_{eq}(x) = N \exp\left(-\frac{V(x)}{k_B T}\right)$$

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

$$\Rightarrow 0 = \left( \frac{\partial}{\partial x} \frac{V'(x)}{m\eta} + k \frac{\partial^2}{\partial x^2} \right) P_{eq}(x)$$

$$\rightarrow P_{eq}(x) = N \exp\left(-\frac{V(x)}{k_B T}\right)$$

$\Rightarrow K = \frac{k_B T}{m\eta}$  Einstein-Sobles-Smolousovski relation (fluctuation-dissipation relation).

Constant external force  $F_0$ :

$$\frac{\partial}{\partial t} P(x,t) = -\frac{F_0}{m\eta} \frac{\partial}{\partial x} P(x,t) + k \frac{\partial^2}{\partial x^2} P(x,t)$$

sometimes called diffusion-advection eq.

(:) Similarity solution with Galilei (Lange) variable  $z = x - \frac{F_0}{m\eta} t$

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

(!!!) Second Einstein relation (linear response):

$$\frac{d}{dt} \langle x(t) \rangle = \frac{F_0}{m\eta} \Rightarrow \langle x(t) \rangle = \frac{F_0}{m\eta} t = \frac{F_0 k}{k_B T} t$$

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

$$\Rightarrow \langle x(t) \rangle_{F_0} = \frac{F_0}{2k_B T} \langle x^2(t) \rangle_0$$

V. The continuous time random walk process

EW Montroll, GH Weiss, J Math Phys 6, 167 (1965).  
 H Sides, Max, Phys Rev B 7, 4502 (1973).  
 MF Silescu, J Stat Phys 10, 421 (1974).  
 H Sides, EW Montroll, Phys Rev B 12, 2455 (1975).  
 J Kleiter, 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:

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

PDF of occurrence of  $i$ th stop at time  $t = t_1 + t_2 + \dots + t_i$

$$\varphi\{\varphi_i(t)\} = \overline{\varphi_i(u)} = \langle e^{-ut} \rangle = \langle e^{-u(t_1+t_2+\dots+t_i)} \rangle$$

independent

$$= \langle e^{-ut_1} \rangle \langle e^{-ut_2} \rangle \dots \langle e^{-ut_i} \rangle = \overline{\varphi_i(u)}$$

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

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

$$\Rightarrow Q_i(u) = \varphi_i(u) \phi(u) = \frac{\varphi_i(u)}{1-\varphi_i(u)}$$

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

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

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

Continuum limit:

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

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

$\Rightarrow p(x, u) = \frac{1}{\sqrt{2\pi u}} \delta(x) + \frac{1}{\sqrt{2\pi u}} \sum_{i=1}^{\infty} \left( p_{i-1}(x) + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} p_{i-1}(x) \right) \frac{1}{\sqrt{2\pi u}} \frac{1}{\sqrt{2\pi u}}$

but from (\*\*):  $\sum_{i=1}^{\infty} p_{i-1}(x) \frac{1}{\sqrt{2\pi u}} = \frac{1}{\sqrt{2\pi u}} \sum_{i=0}^{\infty} p_i(x) \frac{1}{\sqrt{2\pi u}}$

$= p(x, u)$

$\Rightarrow p(x, u) = \frac{1}{\sqrt{2\pi u}} \delta(x) + \frac{1}{\sqrt{2\pi u}} \left( p(x, u) + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} p(x, u) \right)$

$p(x, u) - \delta(x) = \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} p(x, u)$  CTRW with arbitrary  $\varphi(x)$  in 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  $u \approx$  powers  $n(u\tau)^2 \dots$  neglected

Some result for poissonian  $\varphi(t)$ :

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

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

$\langle \tau \rangle = \int_0^{\infty} \tau \varphi(\tau) e^{-u\tau} d\tau = \int_0^{\infty} \tau \frac{d\varphi(u)}{du} \Big|_{u=0} = - \frac{d\langle \tau \rangle}{du} \Big|_{u=0}$   
 $\Rightarrow$  for finite  $\langle \tau \rangle$  we have  $\varphi(u) \approx 1 - \langle \tau \rangle u + \dots$



Up to first order in  $\tau$ :  $\frac{\partial f(u)}{\partial t} = \frac{-1 - u\tau}{1 - u\tau} = \frac{1}{1 - u\tau} - 1$

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

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

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

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

After inverse Laplace transformation:

$$\rho(x, \tau) - \delta(x) = \int_x^0 \frac{\partial^2}{\partial x'^2} \rho(x, \tau') dx'$$

integral form of diffusion eq.

$$\sim \frac{\partial}{\partial x} \rho(x, \tau) = K \frac{\partial^2}{\partial x^2} \rho(x, \tau) \checkmark$$

(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  $\varphi(\tau) \sim \tau^\alpha / \Gamma(1 + \alpha)$ .

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

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  $\rightarrow$  aging, non-ergodicity.

In Laplace space, we find

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

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

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

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 (x):

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

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

This Laplace image defines the Mittag-Leffler function:

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

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

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

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

The Mittag-Leffler function interpolates between an initial stretched exponential law  $\phi \sim \exp(-[t/\tau_k]^\alpha)$  and a terminal inverse power-law.

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

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

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

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:

$$\eta(x, t) = \int_{-\infty}^{\infty} dx' \int_0^t dt' \eta(x', t') \lambda(x-x') \phi(t-t') + \delta(x) \delta(t)$$

initial condition  
needed @  $t=0$

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

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

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

$$P(x, u) = \frac{1 - \phi(u)}{1 - \lambda(x) \phi(u)}$$

(1) Gaussian jump length distribution:

$$\lambda(u) = \exp\left\{-\frac{u^2}{2\sigma^2}\right\} \sim 1 - \frac{u^2}{2\sigma^2}$$

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

$$\Rightarrow P(x, u) = \frac{1}{\tau} = \frac{u\tau + \frac{u^2}{2\sigma^2}}{1} \quad \therefore K = \lambda_{\text{min}} \quad \frac{\sigma^2}{2\tau}$$

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

(2)  $\alpha$ -stable jump length distribution:

$$\lambda(u) = \exp\left(-\frac{|u|^\alpha}{\sigma^\alpha}\right) \sim 1 - \frac{|u|^\alpha}{\sigma^\alpha} \quad \text{as } \langle \delta x^2 \rangle \rightarrow \infty, \quad \lambda(x) \sim \frac{\sigma^\alpha}{|x|^{1+\alpha}}$$

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

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

incomplete  $\gamma$  function

Application to exponential function:

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

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

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

for real-valued  $\mu, \nu$

generalisation through introduction of  $\Gamma$  function:

$$\frac{d^n}{dx^n} x^{m-n} = \frac{(m-n)!}{n!} x^{m-n}$$

for integers  $m, n$

(1.) Phenomenological introduction of fractional order differentiation:

Fractional order differentiation:

fractional calculus.

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

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

"Levy flight"

$$Q(x, t) \sim \frac{t^{-\mu}}{\Gamma(1+\mu)}$$

"Levy flight"

consequences will be drawn

1695 in a letter to GA de l'Hospital, GW Leibniz writes: Thus it follows that  $d^{1/2} x$  will be equal to  $x^{1/2} : x$ , an apparent paradox, from which one may draw