

"Life always has a fat tail"

- E. Fama (Econ Nobel)

Levy Flights - Systematics

1-

(Follows Zaslavsky)

→ Have introduced Levy Flight, L-stable distribution. Here discuss systematically

→ Keep in mind:

= Levy distribution "more general" than Gaussian

- L-stable distributions are Limit Distribution

- Only L-stable dist. of finite variance is Gaussian

→ Levy Flights → generalization of random walks to steps with wild randomness

Levy Distribution

General ideas

(Generalizing CLT)
Limiting Dist.)

$P(x)$ = pdf of random variable x

$\int dx P(x) = 1 \rightarrow$ Normalizability

with

Characteristic fctn:

→ defines moments.

$$\rho(z) = \int_{-\infty}^{+\infty} dx e^{izx} \rho(x)$$

$$\text{so } \langle x^2 \rangle = \lim_{z \rightarrow 0} \left[z^2 \left(\frac{\partial}{\partial z} \right)^2 \rho(z) \right]$$

etc.

a) Stable Distribution:



x_1, x_2 and linear combos $Cx_3 = c_1x_1 + c_2x_2$

ρ is stable if all x_1, x_2, x_3 distr.
according $\rho(x_i)$

Over Gaussians = CLT.

Another class = Levy (1937)

using add. $\textcircled{*}$

$$\rho(x_3)dx_3 = \rho(x_1)\rho(x_2) \delta\left(x_3 - \underbrace{c_1x_1 + c_2x_2}_{C}\right)dx_1dx_2$$

so, generators must satisfy from $\textcircled{*}$

3.

$$\boxed{P(c\varepsilon) = P(c, \varepsilon) P(c_2 \varepsilon)}$$

$$c_1 x_1 + c_2 x_2 = cx$$

on left on left

so

$$\boxed{\ln P(c\varepsilon) = \ln P(c, \varepsilon) + \ln P(c_2 \varepsilon)}$$

These have solution:

$$\ln P_\alpha(c\varepsilon) = (\varepsilon^\alpha)^x = c^\alpha e^{-\frac{\alpha}{2}x(1-\sin\alpha)} / \varepsilon^{\alpha x}$$

where $\left(\frac{c_1}{c}\right)^\alpha + \left(\frac{c_2}{c}\right)^\alpha = 1$.

with arbitrary α .

so $P(x)$ with characteristic function

$$P_\alpha(q) = \exp \left[-c |q|^\alpha \right]$$

$\alpha=2$
Normal

is Levy distribution with Levy index

α

\rightarrow guaranteed

$$P(x) > 0,$$

$$0 < \alpha \leq 2$$

and $\alpha = 2 \rightarrow$ Gaussian

(FT of G is G).

Observe:

$\rightarrow \alpha = 2 \rightarrow$ Gaussian

$\rightarrow \alpha = 1 \rightarrow$ Cauchy

$$p_1 = \frac{1}{\sqrt{\pi}} \left[\frac{1}{x^2 + c^2} \right]$$

\rightarrow large $|x|$, $0 < x < 2$

$$p_2(x) \sim \frac{1}{|x|^{\alpha+1}} \Rightarrow \text{power law}$$

obviously has fat tail for $\alpha < 2$

so

$\rightarrow \langle x^m \rangle$ diverges for ~~$\alpha < 2$~~ $m > \alpha$

c.p. $\alpha < 2 \rightarrow$ Variance diverged
 can't construct $F(x)$

Now consider Lévy process

→ Lévy Process

(analogue/generalization
of diffusion)

- time dependent process
- has Lévy distribution at infinitesimal time Δt .

Consider transition prob:

Chapman-Kolmog. eqn.

$$P(X_0, t_0 | X_N, t_N) = \int dx_1 \int dx_2 \dots \int dx_N [P(x_0, t_0; x_1, t_1) \dots P(x_{N-1}, t_{N-1}; x_N, t_N)]$$

$\underbrace{\qquad\qquad\qquad}_{N}$

$$t_{j+1} - t_j = \Delta t$$

$$N > 1$$

$$t_N - t_0 = N \Delta t$$

and assume process is uniform in space
time (if not?)

$$P(x_j, t_j | x_{j+1}, t_{j+1}) = P(x_{j+1} - x_j; t_{j+1} - t_j) = P(x_{j+1} - x_j, \Delta t)$$

so

$$P(x_N \rightarrow x_0; N \Delta t) = \int dy_1 \dots dy_N P(x_1, \Delta t) \dots P(y_N, \Delta t)$$

Now characteristic

$$P(g) = \int dy_j e^{i g y_j} P(y_j, \Delta t) \rightarrow \text{st}_{\infty}$$

$$P_N(g) = \int_{y^N} dy^N e^{i g y^N} P(y^N, N\Delta t)$$

$$y^N = \sum_j y_j = y_N - y_0$$

(cumulative)

\Rightarrow

$$\boxed{P_N(g) = [P(g)]^N}$$

$$P(g) = P(g | \alpha, c) \quad \underline{\underline{}}$$

$$\{ P(g) = P_x(g, \Delta t) \}$$

$$\{ P_N(g) = P_x(g, c_N) \}$$

above consistent if :

$$c_N = N \Delta t = N \Delta t \frac{\Delta G}{\Delta x} \equiv c N \Delta t = \underline{\underline{c \Delta t}}$$

$\underline{\underline{c \Delta t}}$

$$P_N \rightarrow \left\{ P_\alpha(z, ct) = \exp[-cN\delta + |\varepsilon|^\alpha] \right\}$$

so

$$P_\alpha(\varepsilon, t) = \exp(-ct|\varepsilon|^\alpha)$$

characteristic fctr of Lévy Process

$$P_\alpha(z, t) = \exp(-ct|\varepsilon|^\alpha)$$

2 param: c, α . $\alpha < 2$

$\begin{cases} \alpha=2 \\ \text{diffn.} \end{cases}$

so formally, $P_\alpha(x, t) = \int dz e^{izx} e^{-ct|\varepsilon|^\alpha}$

$$\underset{x \rightarrow \infty}{\sim} t^{1/\alpha+1}$$

and once again see:

$\langle X^2 \rangle \neq \infty$ for $\alpha < 2$, it diverges,

divergent variable.

See Zaburko, Chapt. 15 for mathematical extensions.

Intro to Anomalous Diffusion

Anomalous diffusion:

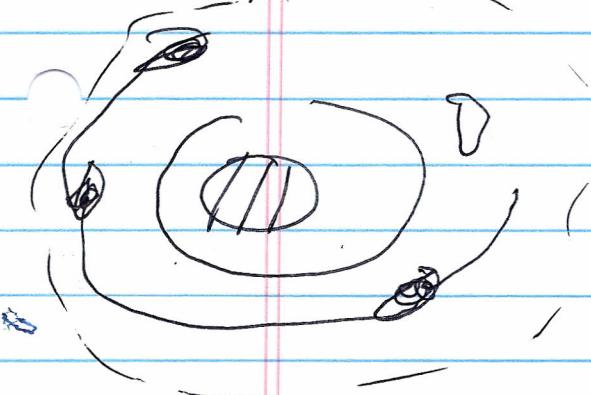
$$\langle \delta r^2 \rangle \sim t^\gamma, \text{ instead } \langle \delta r^2 \rangle \sim t^{1/2}$$

$\gamma > 1 \Rightarrow$ super-diffusive ($\gamma = 2 - \text{ballistic}$)

$\gamma < 1 \Rightarrow$ sub-diffusive

obv. $\gamma \leftarrow H$
connection

Classic example: Weeks, Swinney exp.



rotating tank \rightarrow
eddys

tracer particle:

- trapped many rotation times in eddys

- occasional large flights

Anomalous diffusion can manifest often
oddities \rightarrow up-gradient transport (I)

Q: How derive kinetic equation?

Review F-P:

$$-\frac{\partial}{\partial x} \cdot \vec{P} = \frac{\Delta \vec{x}}{\Delta t}$$

Now, how represent anomalous diffusion?

- CTRW (Continuous Time Random Walk)
- Fractional kinetics * (main emphasis)

→ CTRW

- to represent allow large steps

$$\text{e.g. } P(\Delta z) \sim 1/\Delta z^{1+\alpha}$$

but $\alpha < 2$ precludes use of F-P expansion.

i.e. can't simply use fat tail distribution

- CTRW \Rightarrow accommodate flights/sticking in random walk model

\rightarrow release time from dummy role \Rightarrow allow evolve dynamically so walker position does.

\rightarrow allow large (divergent) steps in space to be accomplished in large step time so variance finite.

⇒ CTRW

→ position of n^{th} step

$$r_n = r_0 + \Delta r_1 + \dots + \Delta r_{n-1} + \Delta r_n$$

$$t_n = t_0 + \Delta t_1 + \Delta t_2 + \dots + \Delta t_n$$



time of n^{th} step

| Need $P(\Delta t_i)$

2 approaches to CTRW (see later review)

→ waiting model:

- steps in position, time independent

- need specify 2 probabilities

$\begin{cases} \Delta r \\ \Delta t \end{cases}$

idea: particle waits Δt in position (sticking), then jumps Δr on no time.

→ velocity model

- $\Delta t \rightarrow$ traveling time of particle

i.e.

11.

$$\Delta t = |\Delta r| / v \quad \hookrightarrow \text{const } v.$$

so increments satisfy $f(\Delta t - |\Delta r| / v) P(\Delta r)$

→ general → specify joint prob

Now to build up CTRW equations:
extended Chapman-Kolmogorov Eqn,

for dist jump pts:

chap-kolma

step.

$$Q(z, t) = \int d\Delta z \int d\Delta t Q(z - \Delta z, t - \Delta t) P(\Delta z, \Delta t)$$

+ i.c. + so.

so P ≡ probability of walker to be at position z at time t ~~need specify~~

① in waiting model:

$$P(z, t) = \boxed{P(\Delta z, \Delta t) = P(\Delta z) P(\Delta t)}$$

$$P_w(z, t) = \int d(\Delta t) Q(z + \Delta t) P_w(\Delta t)$$



$$\Phi_w(\Delta t) = \int_{\Delta t}^{\infty} dt' P_{\Delta t}(t' + \Delta t)$$

prob to wait at least Δt

(2)
velocity model:

$$P(\Delta z, \Delta t) = f((\Delta t - |\Delta z|)/v) \rho(\Delta z)$$

$$P(z, t) = \int_{-vt}^{vt} d(\Delta z) \int_{\Delta t}^{+} d(\Delta t) Q(z - \Delta z, t - \Delta t) \Phi_v(\Delta z, \Delta t)$$

with:

$$\Phi_v = \frac{1}{2} \int (1_{|\Delta z|} - v \Delta t) \int_{\Delta z}^{\infty} dz' \int_{\Delta t}^{+} dt' \delta(t' - |z'|/v) \cancel{P(z')}$$

prob. to make step of at least length $|\Delta z|$ and duration Δt

key question \rightarrow what does the mes look like??

CTRW "diffusion"

→ depends on distribution of step increments, in both time, space
 - contrast F&F Eq.

→ small steps both recover "normal" diffusion

but

- { small spatial steps
 Levy distributed (long) waiting times

⇒ subdiffusion, sticking

- { small time steps
 Levy distributed (long) spatial steps

⇒ superdiffusion

→ CTRW eqns. Non-local, non-Markovian
 in space + time

C-C

$$\nabla \nabla_x^2 P \rightarrow \partial_x \left\{ \int dx' \int dt H \right. *$$

$$R(x-x_j^t + t) \partial_x P(x_j^t + t)$$

↑
scale of kernel
(critical) → spatial approach
via factorization

How solve this mess?

→ leave long tail, RCR
unstructured
in cheap form

→ Fourier-Laplace Transform = i.e. work
with generating function.

Recall: generating fctn.

$$P(k) = \int e^{-ik\Delta z} P(\Delta z) d\Delta z$$

\approx

$$i^n \sum_{k=0}^{\infty} \partial_k^n P(k) \equiv \int dz (\Delta z)^n P(\Delta z)$$

$$i^n \left[\partial_k^n P(k) \right]_{k=0} = \langle (\Delta z)^n \rangle$$

For CTRW solution, concentrate on asymptotic, large $|z|$ regime \Leftrightarrow low k

Make Taylor expansion:

$$\bullet P(k) = p(0) + k \partial_k p(0) + \frac{k^3}{2} \partial_k^2 p(0)$$

\Rightarrow

$$P(k) = 1 - \cancel{\langle (\Delta z) \rangle} k - \frac{1}{2} \langle (\Delta z)^2 \rangle k^2 + \dots$$

\curvearrowleft
symmetry

$$P(k) = 1 - \frac{1}{2} \langle (\Delta z)^2 \rangle k^2$$

Similarly, time distributions
(which can be 1 sided)

$$P(s) = 1 - \cancel{\langle t \rangle} s + \dots$$

\curvearrowleft analogy

which brings us back to Levy,

$$\cancel{P^{L^\alpha}(k) = \exp[-\alpha|k|^\alpha]}$$

$$\approx 1 - \alpha|k|^\alpha + \dots$$

and proceed following posted materials.

and proceed at at notes.

Summary

→ CTRW: → solve FA

- dist. $\Delta t, \Delta z$
 $\xrightarrow{\text{new}} \text{stochastic}$

- works directly with integral eqn
 $\xrightarrow{\text{Chap - Kolmog}}$

- factorizes $P(\Delta z, \Delta t)$

- needs fast fail $P(\Delta t)$ dist.