

Anomalous Diffusion



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Diffusion Equation

Langevin Equation:

$$\left\{ \begin{array}{l} \frac{dx}{dt} = v \\ \frac{dv}{dt} = -\frac{1}{m} \frac{dV(x)}{dx} - \gamma v + \xi(t) \end{array} \right.$$

$$\langle \xi(t) \xi(s) \rangle = \frac{2k_B T \gamma}{M} \delta(t-s)$$

White Noise

Fluctuation
Dissipation

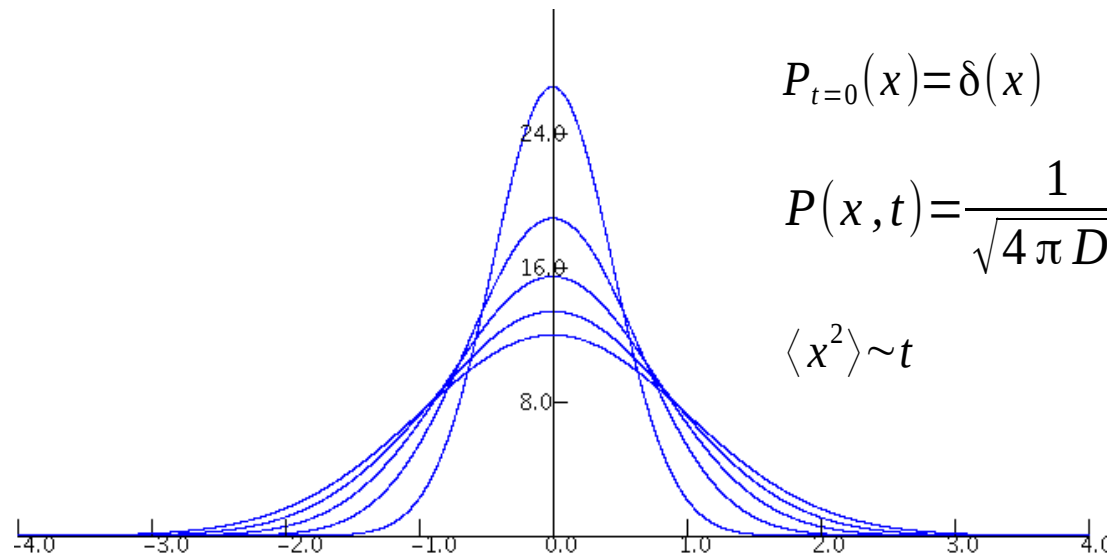
$$\gamma \dot{x} = \xi$$

- Free
- Overdamped

Fokker-Planck Equation:

$$\frac{\partial P(x,t)}{\partial t} = D_0 \frac{\partial^2 P}{\partial x^2}$$

$$D_0 = \frac{k_B T}{m \gamma}$$



$$P_{t=0}(x) = \delta(x)$$

$$P(x,t) = \frac{1}{\sqrt{4\pi D_0 t}} e^{-\frac{x^2}{4D_0 t}}$$

$$\langle x^2 \rangle \sim t$$

Central Limit Theorem

$$X_i \text{ i.i.d., } \begin{array}{l} \text{average}(X_i) = \mu \\ \text{variance}(X_i) = \sigma^2 \end{array} \quad S_n = \sum_{i=1}^n X_i \quad S_n \stackrel{n \rightarrow \infty}{\sim} N(n\mu, n\sigma^2)$$

$$\varphi(u) = L\{f(t)\} := \int_0^\infty e^{-ut} f(t) dt$$

$$f(t) * g(t) := \int_0^t f(t-\tau) g(\tau) d\tau = g(t) * f(t)$$

$$N(a_1^2, b_1) * N(a_2^2, b_2) = N(a_1^2 + a_2^2, b_1 + b_2)$$

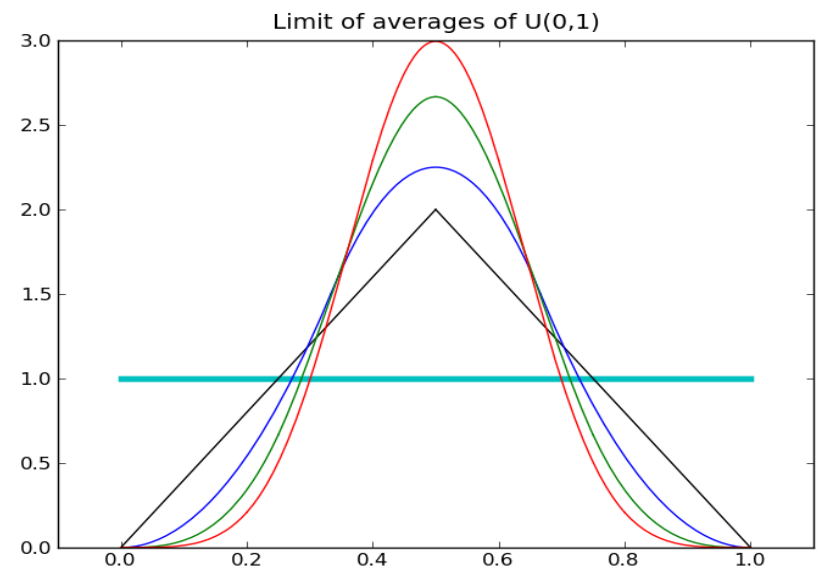
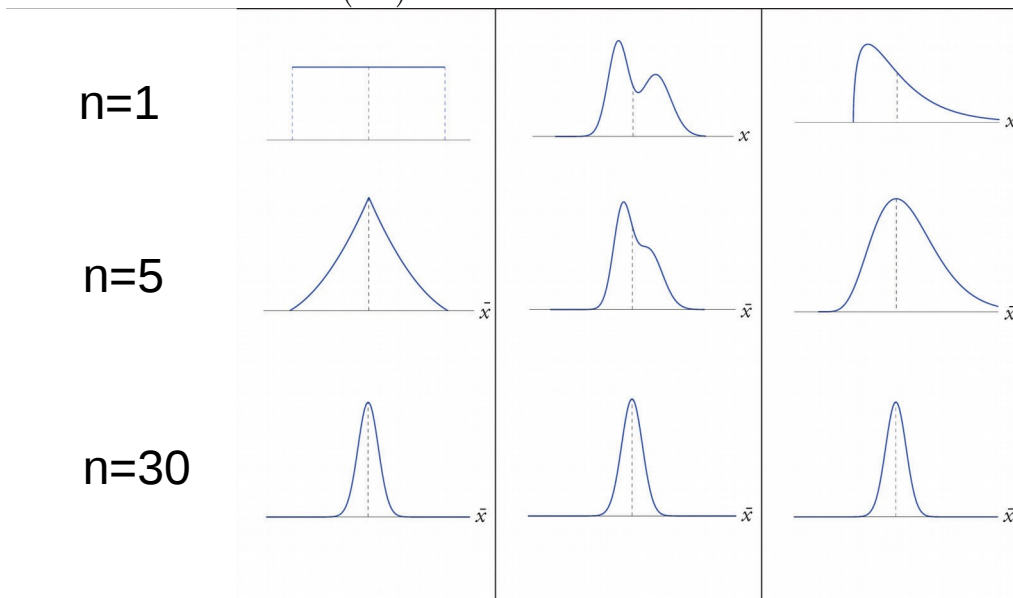
$$X_1 \sim f \quad X_2 \sim g$$

$$X_1 + X_2 \sim f * g$$

$$\varphi_{f * g}(u) = \varphi_f(u) \varphi_g(u)$$



$P(X)^{*n}$



Fractional Brownian Motion

Continuous Time Gaussian Process:
$$\begin{cases} E[B_H(t)] = 0 \\ E[B_H(t)B_H(s)] = \frac{1}{2}(|t|^{2H} + |s|^{2H} - |t-s|^{2H}) \end{cases}$$

$H > 1/2$: Increments positively correlated

$H \in (0, 1)$

$H = 1/2$: Brownian Motion

Hurst Index

$H < 1/2$: Increments negatively correlated

The following integral satisfies the definition

$$B_H(t) = \frac{1}{C(H)} \int_{-\infty}^0 [(t-s)^{H-1/2} - (-s)^{H-1/2}] dB(s) + \int_0^t [(t-s)^{H-1/2}] dB(s)$$

with $B(t)$ Brownian Motion

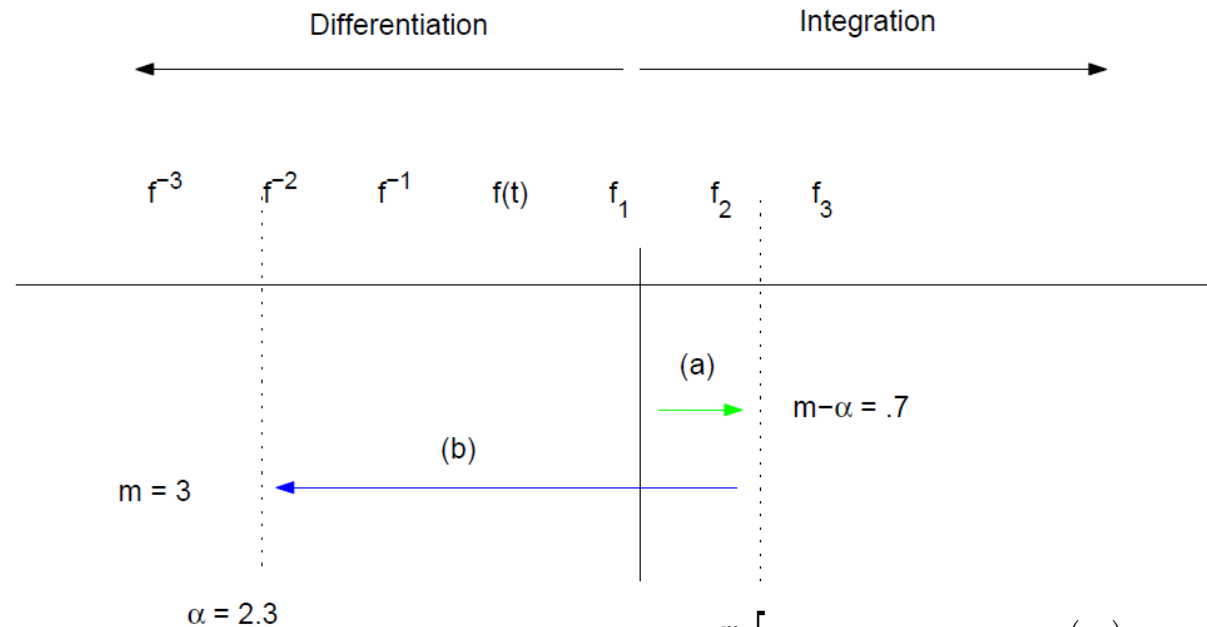
$$B_H(t) = \frac{\Gamma(H+1/2)}{C(H)} (I^{H-1/2}) B(t)$$

I : differintegral operator

Fractional Integral & Fractional Derivative (1)

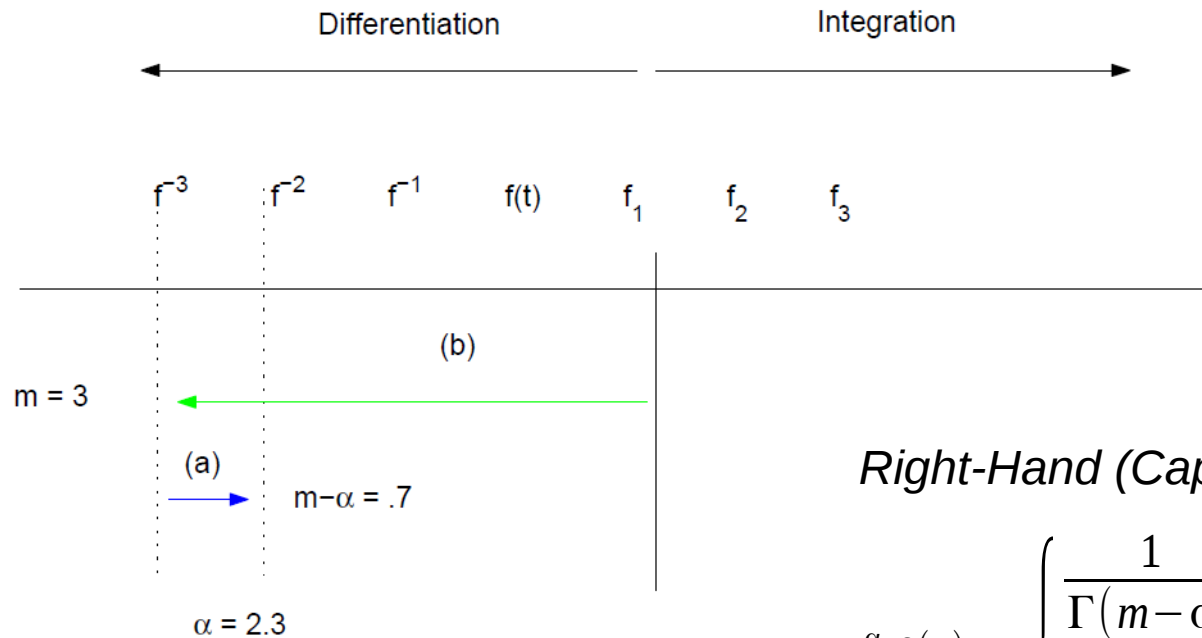
$$\int_a^x d\sigma_1 \int_a^{\sigma_1} d\sigma_2 \dots \int_a^{\sigma_{n-1}} d\sigma_n f(\sigma_n) = \frac{1}{(n-1)!} \int_0^t (t-\tau)^{n-1} f(\tau) d\tau$$

$$I^\alpha f(t) = \frac{1}{\Gamma(\alpha)} \int_0^t (t-\tau)^{\alpha-1} f(\tau) d\tau$$



Left-Hand fractional Derivative:
$$D^\alpha f(t) := \begin{cases} \frac{d^m}{dt^m} \left[\frac{1}{\Gamma(m-\alpha)} \int_0^t \frac{f(\tau) d\tau}{(t-\tau)^{\alpha+1-m}} \right], m-1 < \alpha < m \\ \frac{d^m}{dt^m} f(t), \alpha = m \end{cases}$$

Fractional Integral & Fractional Derivative (2)



Right-Hand (Caputo) fractional Derivative:

$$D_*^\alpha f(t) := \begin{cases} \frac{1}{\Gamma(m-\alpha)} \int_0^t \frac{f^{(m)}(\tau) d\tau}{(t-\tau)^{\alpha+1-m}}, & m-1 < \alpha < m \\ \frac{d^m}{dt^m} f(t), & \alpha = m \end{cases}$$

- If C is a constant $D_*^\alpha C = 0$ $D^\alpha C \neq 0$
- For RH Fractional Differential Equation we don't need fractional order initial conditions

Mittag-Leffler function

$$E_\alpha(-\lambda t^\alpha) = \sum_{k=0}^{\infty} \frac{(-\lambda t^\alpha)^k}{\Gamma(\alpha k + 1)}$$

Colored noise

$$M \ddot{x}(t) + M \int_0^t \gamma(t-\tau) \dot{x}(\tau) d\tau + U'(x) = \xi \quad \text{Langevin equation with friction kernel}$$

$$\langle \xi(t) \xi(s) \rangle = K(t-s) \quad \text{Colored noise}$$

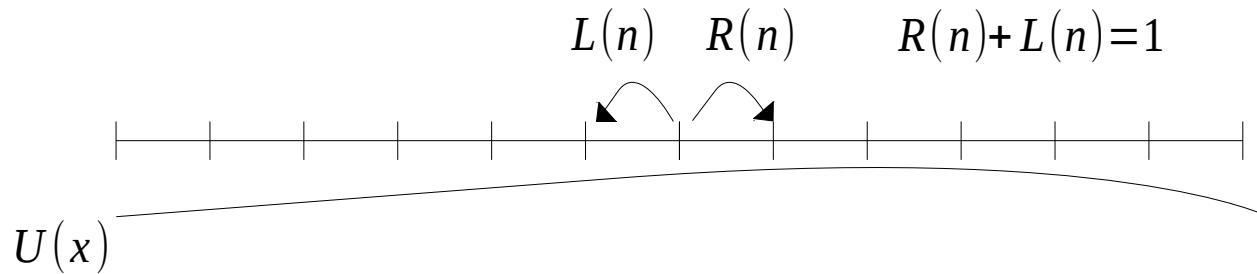
$$M k_B T \gamma(t-s) = K(t-s) \quad \text{Generalized Fluctuation-Dissipation}$$

$$M \ddot{x}(t) + M \gamma_\alpha \frac{\partial^{\alpha-1}}{\partial t^{\alpha-1}} \dot{x}(t) + U'(x) = \xi \quad \text{Fractional form}$$

$$K(t-s) \sim |t-s|^{-\alpha} \xrightarrow[\alpha \neq 1]{0 \leq \alpha \leq 2} C_\nu(t) = E_{2-\alpha}(-\text{const } t^{2-\alpha}) \longrightarrow \text{MSD} = \langle x^2 \rangle \sim t^\alpha$$

$$\left\{ \begin{array}{l} \frac{\partial P_{fBM}(x,t)}{\partial t} = \alpha D_0 t^{\alpha-1} \frac{\partial^2 P_{fBM}(x,t)}{\partial x^2} \\ P_{t=0}(x) = \delta(x) \end{array} \right. \longrightarrow \boxed{P(x,t) = \frac{1}{\sqrt{4\pi D_0 t^\alpha}} e^{-\frac{x^2}{4D_0 t^\alpha}}}$$

Continuum Limit of CTRWs (1)



$\psi(\tau)$ = Waiting time distribution

$p_i(n)$ = Probability to be in site n after i jumps

$p_{i+1}(n) = R(n-1)p_i(n-1) + L(n+1)p_i(n+1)$

$$P(n, t) = \sum_{i=0}^{\infty} p_i(n) Q_i(t)$$

$Q_i(t)$ = Probability of i jumps until time t

$p_i(n) \longleftrightarrow p_i(x)$ Continuum limit

Taylor Expansion

$$R(x-a)p_i(x-a) = R(x)p_i(x) + \frac{\partial}{\partial x} [R(x)p_i(x)] + \frac{\partial^2}{\partial x^2} [R(x)p_i(x)] \frac{a^2}{2} + \dots$$

Detailed Balance

$$R(x) - L(x) \simeq \frac{a F(x)}{2k_B T} \quad R(x) \simeq L(x) \simeq \frac{1}{2} : F(x) \text{ slowly varying}$$

Continuum Limit of CTRWs (2)

If the first moment of $\psi(\tau)$ exists

$$\underbrace{\frac{p_{i+1}(x) - p_i(x)}{\langle \tau \rangle}}_{\frac{\partial p(x,t)}{\partial t}} = \underbrace{\frac{a^2}{2\langle \tau \rangle}}_{K_1 = \lim_{\substack{\langle \tau \rangle \rightarrow 0 \\ a^2 \rightarrow 0}} \frac{a^2}{2\langle \tau \rangle} = D_0} \left[\frac{\partial^2}{\partial x^2} p_i(x) - \frac{\partial}{\partial x} \left(\frac{F(x)}{k_B T} p_i(x) \right) \right] \longrightarrow \text{Normal Fokker-Planck}$$

What about $\psi(\tau) \stackrel{\tau \text{ large}}{\sim} \frac{\alpha A_\alpha}{\Gamma(1-\alpha) \tau^{1+\alpha}} \quad \alpha < 1 \quad \longrightarrow \quad \langle \tau \rangle \text{ diverges}$

$\hat{\psi}(u) = 1 - A_\alpha u^\alpha + c_1 (A_\alpha u^\alpha)^2 + \dots \longrightarrow u^\alpha \hat{P}(x, u) = K_\alpha L_{fp} \hat{P}(x, u)$

$$K_\alpha = \lim_{\substack{A_\alpha \rightarrow 0 \\ a^2 \rightarrow 0}} \frac{a^2}{2A_\alpha}$$

Generalized Diffusion Constant

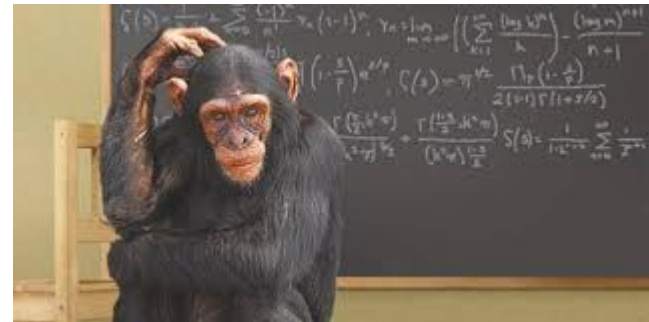
$$[K_\alpha] = \left[\frac{m^2}{\text{sec}^\alpha} \right]$$

Fractional Fokker-Planck Equation

$$D_*^\alpha P(x, t) = K_\alpha \left[\frac{\partial^2}{\partial x^2} - \frac{\partial}{\partial x} \left(\frac{F(x)}{k_B T} \right) \right] P(x, t)$$

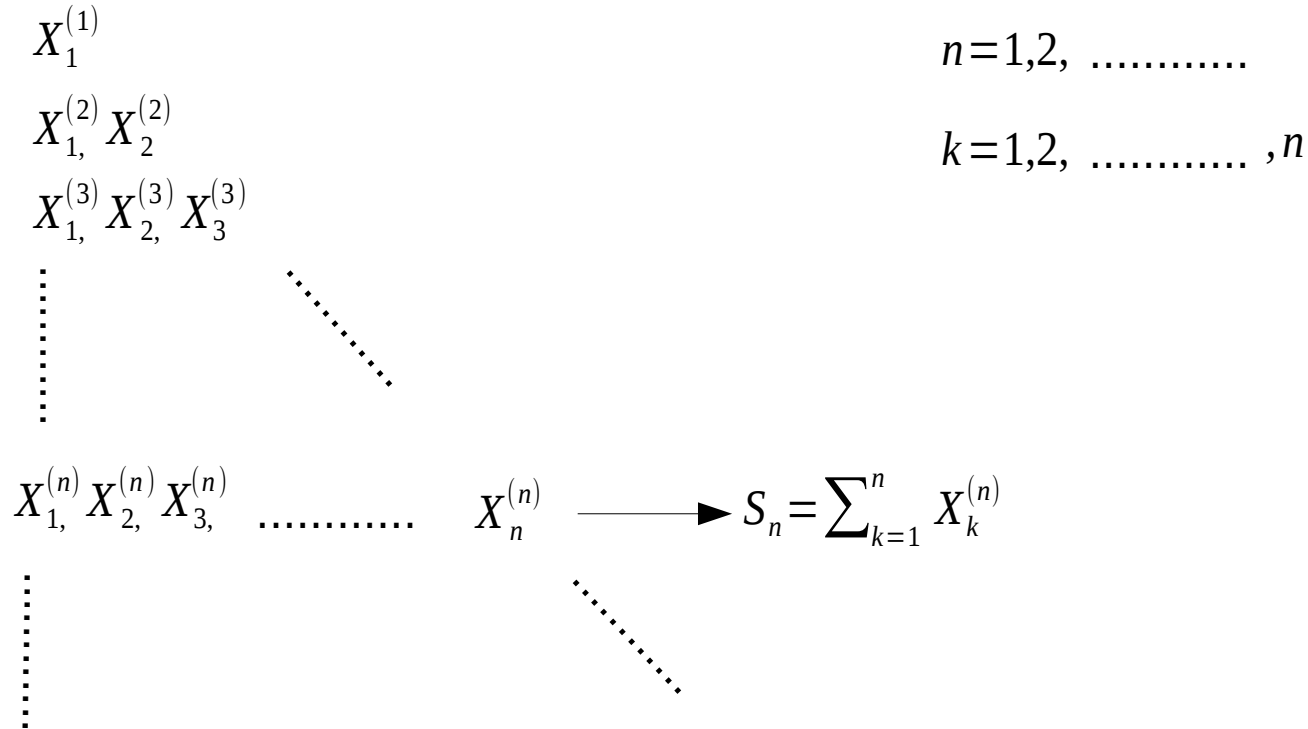
$MSD = \langle x^2 \rangle \sim t^\alpha \quad \alpha < 1 \quad \text{Subdiffusion}$

Thank you & Merry Christmas !



....of the monkey year!

Backup slide 1: Central Limit Theorem (2)



- In every row $n \in \mathbb{N}$ the random variables $X_1^{(n)}, X_2^{(n)}, X_3^{(n)}, \dots, X_n^{(n)}$ are independent
- The $X_k^{(n)}$ are uniformly, asymptotically negligible, namely $\max_k P\{|X_k^{(n)}| \geq \epsilon\} \xrightarrow{n \rightarrow \infty} 0, \forall \epsilon > 0$

Backup slide 2: Central Limit Theorem (3)

CLP1: $X_1^{(n)}, X_2^{(n)}, \dots, X_n^{(n)}$ independent:

Infinitely divisible laws

$$X \sim \sum_{k=1}^n X_k^{(n)} \quad \varphi(u) = (\varphi_n(u))^n$$

CLP2: $X_1^{(n)}, X_2^{(n)}, \dots, X_n^{(n)}$ independent +

Self-Decomposable laws

$$\exists a_n, b_n \text{ and } X_k \text{ s.t. } X_k^{(n)} = \frac{1}{a_n} \left(X_k - \frac{b_n}{n} \right):$$

$$X \sim aX' + Y_a \quad \varphi(u) = \varphi(au)\varphi_a(u)$$

$$a \in (0,1) \quad X' \sim X$$

CLP3: $X_1^{(n)}, X_2^{(n)}, \dots, X_n^{(n)}$ i.i.d.:

Stable laws

$$c_1 X_1 + c_2 X_2 \sim aX + b$$

$$e^{ibu} \varphi(au) = \varphi(c_1 u) \varphi_a(c_2 u)$$

$$\text{Symmetric stable laws: } \varphi(u) = e^{-b|u|^\alpha}$$

Backup slide 3: Central Limit Theorem (4)

Example of construction of a triangular array

$$P(\lambda) = \frac{\lambda^k e^{-\lambda}}{k!} \quad \text{Poisson law} \quad \longrightarrow \quad \varphi_P(u) = e^{\lambda(e^{iu} - 1)}$$

$$B(k; N, p) = \binom{N}{k} p^k (1-p)^{N-k} \quad \text{Binomial law} \quad \longrightarrow \quad \varphi_B(u) = (1-p + pe^{it})^N$$

$$\text{Triangular array: } X_k^{(n)} \sim B(N=1, p = \frac{\lambda}{n})$$

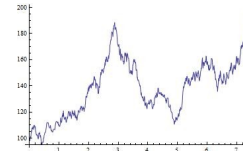
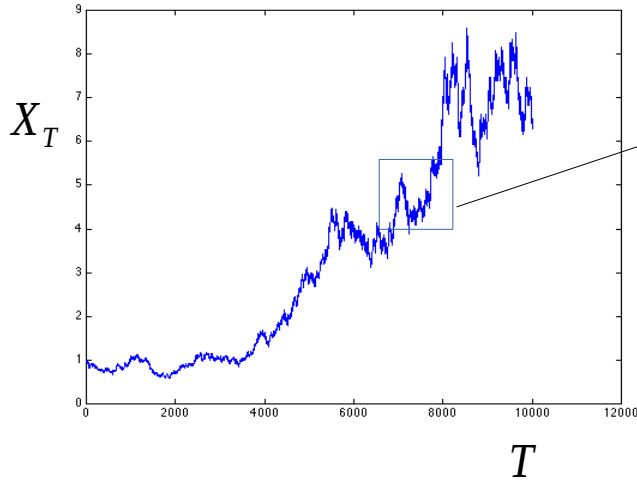
$$S_n = X_1^{(n)} + X_2^{(n)} + \dots + X_n^{(n)} \sim B(n, \frac{\lambda}{n}) \xrightarrow{n \rightarrow \infty} P(\lambda)$$

$$X_1^{(1)} \sim B(1, \lambda) \longrightarrow \varphi(u) = (1 - \lambda + \lambda e^{it}) \xrightarrow[\text{rescaling}]{\text{centering}} \varphi(u) = e^{ibt} (1 - \lambda + \lambda e^{iat})$$

$$X_1^{(2)} \sim B(1, \frac{\lambda}{2}) \longrightarrow \varphi(u) = \left(1 - \frac{\lambda}{2} + \frac{\lambda}{2} e^{it}\right)$$

~~No a and b make the job~~

Backup slide 4: Stationary and Self Similar Process (4)



Process defined through the law of its increments

Levy process:

- $X_{t=0} = 0$
- *Independent and Stationary Increments*
- *Continuity in probability*

$$X_t = \sum_{k=1}^n X_{\frac{kt}{n}} - X_{\frac{(k-1)t}{n}}$$

ϕ : law of the increment

φ : law of X_T

Stationary increment: $\phi_{s,t} = \phi_{\tau=t-s}$

Stationary process from an infinitely divisible law:

$$\phi_t(u) = (\varphi(u))^{t/T}$$

iff φ is infinitely divisible

Self-similar process: $\phi_{as,at}(u) = \phi_{s,t}(bu)$

Self-similar process from a self-decomposable law:

$$\phi_{s,t}(u) = \frac{\varphi\left(\left(\frac{t}{T}\right)^H u\right)}{\varphi\left(\left(\frac{s}{T}\right)^H u\right)}$$

iff φ is self-decomposable
 $b = a^H$