

MTH 9831

HW10

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HW 10

(1) $X \sim T(\alpha, \lambda)$ $f_X(x) = \frac{\lambda^\alpha}{\Gamma(\alpha)} e^{-\lambda x} x^{\alpha-1} \mathbb{1}_{[0, \infty)}(x)$

C.H.F.

$$\begin{aligned} \phi_X(u) &= E(e^{iux}) = \int_{-\infty}^{\infty} e^{iux} \frac{\lambda^\alpha}{\Gamma(\alpha)} e^{-\lambda x} x^{\alpha-1} \mathbb{1}_{[0, \infty)}(x) dx \\ &= \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^{\infty} e^{-(\lambda - iu)x} x^{\alpha-1} dx \end{aligned}$$

Let $v = (\lambda - iu)x$ ~~$x = \frac{v}{\lambda - iu}$~~

~~$$= \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^{\infty} e^{-v} \left(\frac{1}{\lambda - iu} \right)^{\alpha-1} \frac{1}{\lambda - iu} d\left(\frac{1}{\lambda - iu} v \right)$$~~

~~$$= \left(\frac{\lambda^\alpha}{\lambda - iu} \right) \int_0^{\infty} \frac{1}{\Gamma(\alpha)} e^{-v} v^{\alpha-1} dv$$~~

$$= \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^{\infty} e^{-v} x^{\alpha-1} dx$$

$$= \frac{\lambda^\alpha}{\lambda^\alpha} \int_0^{\infty} \frac{v^\alpha}{\Gamma(\alpha)} e^{-v} v^{\alpha-1} dv$$

$$= \left(\frac{\lambda}{\lambda} \right)^\alpha = \left(\frac{\lambda}{\lambda - iu} \right)^\alpha$$

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For infinitely divisible.

Suppose $X \sim \Gamma(n, \lambda)$ is a sum of n independent $\Gamma(1, \lambda) \equiv \text{EXP}(\lambda)$

So, there exists a sequence of i.i.d random variables $\{X_i\}_{i=1}^n$.

$$\text{C.H.F} = \left(\frac{\lambda}{\lambda - iu}\right)^n = (\phi_{X_1}(u))^n \quad X_i \sim \text{EXP}(\lambda)$$

(b) Frullani integral

$$\left(\frac{\lambda}{\lambda - z}\right)^a = e^{\int_0^\infty (e^{zx} - 1) \frac{\partial}{\partial x} e^{-\lambda x} dx}$$

Plug $z = iu$ in the integral.

$$\phi_X(u) = \left(\frac{\lambda}{\lambda - iu}\right)^a = e^{\int_0^\infty (e^{iux} - 1) \frac{\partial}{\partial x} e^{-\lambda x} dx}$$

From Lévy-Khintchine representation:

$$\phi_X(u) = E[e^{iux}] = e^{i\mu u - \frac{\sigma^2}{2} u^2 + \int (e^{iux} - 1 - iux \mathbb{I}_{|x| < 1}) d\nu(x)}$$

Compare the equations.

$$\text{we get } (\mu = 0, \sigma = 0 \quad \nu(dx) = \frac{\partial}{\partial x} e^{-\lambda x} \mathbb{I}_{(0, \infty)}(x) dx)$$

To show this.

$$\int x \mathbb{I}_{\{0 < |x| \leq 1\}} v(dx) \\ = \int_0^1 \frac{2}{x} \cdot e^{-\lambda x} dx = \frac{2}{\lambda} (1 - e^{-\lambda})$$

$$\text{Let } \mu_0 = \mu - \int x \mathbb{I}_{\{|x| \leq 1\}} v(dx) = 0.$$

Characteristic exponent:

$$\Phi(u) = iu\mu_0 - \frac{\sigma^2}{2}u^2 + \int (e^{iux} - 1) dv(x)$$

$$v(dx) = \frac{2}{x} e^{-\lambda x} \mathbb{I}_{(0, \infty)}(x)$$

$$(c) \text{ Levy triplet. } \mu_0 = 0, \sigma = 0, dv = \frac{2}{x} e^{-\lambda x} \mathbb{I}_{(0, \infty)}(x) dx$$

$$(d) \varphi_{x_1, -x_2}(u) = \varphi_{x_1}(u) \cdot \varphi_{-x_2}(u)$$

$$= \left(\frac{\lambda_1}{\lambda_1 - iu} \right)^{\alpha_1} \cdot \left(\frac{\lambda_2}{\lambda_2 + iu} \right)^{\alpha_2}$$

$$= \exp \left[\int_0^\infty \left[(e^{iux} - 1) \cdot \frac{\alpha_1}{x} e^{-\lambda_1 x} + (e^{-iux} - 1) \frac{\alpha_2}{x} e^{-\lambda_2 x} \right] dx \right]$$

$$\text{Exponent } \Phi_{x_1, -x_2}(u) = \int_0^\infty (e^{iux} - 1) \left(\frac{\alpha_1}{x} e^{-\lambda_1 x} - e^{-iux} \frac{\alpha_2}{x} e^{-\lambda_2 x} \right) dx$$

$$v(dx) = \left(\frac{\alpha_1}{x} e^{-\lambda_1 x} - e^{iux} \frac{\alpha_2}{x} e^{-\lambda_2 x} \right) \cdot dx$$

(2) $Y = X_T$ $X_t \sim \text{Levy}(\mu, \sigma^2, \nu)$

$T \sim \text{EXP}(\lambda)$ $f_T(t) = \lambda e^{-\lambda t} \mathbb{I}_{(0, \infty)}(t)$

Y is infinite divisible, there exist some i.i.d. $\& Y_j, j=1, 2, \dots, n$ s.t

$$\varphi_n(u) = (\varphi_{Y_j}(u))^n$$

For Fixed t .

$$E(e^{iuX_t}) = \varphi_{X_t}(u) = (\varphi_{X_1}(u))^t = E(e^{iuX_1})^t.$$

But for random T

$$E(e^{iuX_T}) = \varphi_{X_T}(u) \neq (\varphi_{X_1}(u))^T$$

Because $\varphi_{X_T}(u)$ is a deterministic function in u , and $(\varphi_{X_1}(u))^T$ is a random ~~variab~~ ^{func} function

However, we can say,

$$E(e^{iuX_T} | T=t) = E(e^{iuX_t})$$

$$\text{Then, } \varphi_{X_T}(u) = E(e^{iuX_T}) = \int_0^\infty E(e^{iuX_T} | T=t) f_T(t) dt$$

$$= \lambda \int_0^\infty (\varphi_{X_1}(u))^t e^{-\lambda t} dt. \quad (\text{let } \varphi_{X_1}(u) = e^{\Phi(u)})$$

$$= \lambda \int_0^\infty e^{-(\lambda - \Phi(u))t} dt$$

$$= \frac{\lambda}{\lambda - \Phi(u)}$$

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(3)

(a) Time: $Z_t \sim \text{Poisson}(\lambda)$ process

So, $Z_t \sim \text{Levy}(0, 0, \lambda \delta(x-1)dx)$

Space: $W_t \sim \text{BM}$

$W_t \sim \text{Levy}(0, 1, 0)$ process

~~So~~ So, subordinated process

$Y_t = W_{Z_t}$ (Time changed B.M.)

We can see Y_t is flat until Z_t jumps.

$$Z_t = \begin{cases} 0 & \text{until 1st jump} \\ 1 & \text{until 2nd jump} \\ 2 & \text{until 3rd jump} \\ \vdots & \end{cases}$$

$$W_{Z_t} = \begin{cases} 0 & \text{until } Z_t = 1 \\ W_1 & \text{until } Z_t = 2 \\ W_2 & \text{until } Z_t = 3 \\ \vdots & \end{cases}$$

Its jump size on the k^{th} jump is

$$X_k - X_{k-1} \sim N(0, 1)$$

Therefore, W_{Z_t} is a compound poisson(λ) with ~~independen~~ independent $N(0, 1)$ jumps

(b) Let $T_i \sim \text{Exp}(\lambda)$ is Z 's jump times.

$$V_j \sim \text{Exp}(\alpha) \quad j = 1, 1.$$

$$Z_t = \begin{cases} 0 & \text{until } T_1 \\ V_1 & \text{until } T_1 + T_2 \\ V_2 & \text{until } T_1 + T_2 + T_3 \\ \vdots & \end{cases}$$

$$W_{Z_t} = \begin{cases} 0 & \text{until } Z_t = V_1 \\ W_{V_1} & \text{until } Z_t = V_1 + V_2 \\ W_{V_1+V_2} & \text{until } Z_t = V_1 + V_2 + V_3 \\ W_{V_1+V_2+V_3} & \text{until } Z_t = V_1 + V_2 + V_3 + V_4 \end{cases} \begin{array}{l} T_1 \\ T_1 + T_2 \\ T_1 + T_2 + T_3 \\ T_1 + T_2 + T_3 + T_4 \end{array}$$

W_{Z_t} is a pure jump process. its jump size on the k^{th} jump at time $\sum_{j=1}^k T_j$ is $N(0, V_k)$

We need to find $\int_0^{\infty} dF_{X_s}(\alpha) p(ds)$

$$dF_{X_s}(x) = f_{X_s}(x) dx$$

$$dp(s) = p(ds) = p(s) ds \quad \text{as long as } dp \ll ds$$

We need to find $f_{X_s}(x)$, pdf of $N(0, \text{Exp}(\alpha))$ in this case.

If $V_k \sim \exp(\lambda)$ is given, we can do it.

Say $V_k = v$, then the pdf is

$$f(x, v) = \frac{1}{\sqrt{2\pi v}} e^{-\frac{x^2}{2v}}$$

Therefore,

$$\begin{aligned} E(f(x, v)) &= E(E(f(x, v) | V = v)) \\ &= \int_{-\infty}^{+\infty} f(x, v) g(v) dv \quad (g(v) \text{ is pdf of } \exp(\lambda)) \\ &= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi v}} e^{-\frac{x^2}{2v}} \cdot \lambda e^{-\lambda v} dv \\ &= \frac{\lambda}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \frac{1}{\sqrt{v}} e^{-\lambda v - \frac{x^2}{2v}} dv. \end{aligned}$$

(4)

(a) Let $f = \ln S_t$.

$$df = f_t dt + f_s dS^C + \frac{1}{2} f_{ss} d[S^C]_t + f(t, S_t) - f(t, S_{t-})$$

$$dS_t^C = S_t \cdot dX_t^C \quad X_t^C = \mu t + \sigma W_t.$$

$$\ln S_t - \ln S_0 = \frac{1}{S_t} dS_t^C + \frac{1}{2} \frac{1}{S_t^2} d[S^C]_t + \sum_{0 < s < t} [\ln S_s - \ln S_{s-}]$$

$$\begin{aligned} \ln S_t &= \ln S_0 + \frac{S_t}{S_t} dX_t^C + \frac{1}{2} \frac{S_t^2}{S_t^2} d[X_t^C]_t + \sum_{0 < s < t} [\ln S_s - \ln S_{s-}] \\ &= \ln S_0 + \frac{S_t}{S_t} (\mu dt + \sigma dW_t) - \frac{S_t^2}{2S_t^2} \cdot \sigma^2 dt + \sum_{0 < s < t} [\ln S_s - \ln S_{s-}] \\ &= \ln S_0 + \left(\frac{S_t}{S_t + \Delta S} \left(\mu dt + \sigma dW_t \right) - \frac{S_t^2}{2(S_t + \Delta S)^2} \right) dt \\ &\quad + \frac{S_t}{S_t + \Delta S} \cdot \sigma \cdot dW_t + \sum_{0 < s < t} [\ln S_s - \ln S_{s-}] \end{aligned}$$