

$\left\{ \begin{array}{l} \text{MGF} \rightarrow \text{Laplace transform} \quad \mathbb{E} e^{tx} \\ \text{c.f.} \rightarrow \text{Fourier transform} \quad \mathbb{E} e^{itx} \end{array} \right.$

$\left\{ \begin{array}{l} \text{real exponential} \rightarrow \text{monotone, explode/vanish} \\ \text{complex exponential} \rightarrow \text{periodic} \end{array} \right.$

c.f. has 1-to-1 correspondence with dist
from Levy's inversion formula (inverse Fourier)

good tool to deal with sum of independent
r.v.

5.7.7: X_1, \dots, X_n independent, $X_i \sim N(\mu_i, 1)$,

$Y = X_1^2 + \dots + X_n^2$, calculate c.f. of Y .

pf:

$$\phi_Y(t) = \prod_{j=1}^n \phi_{X_j^2}(t), \text{ and}$$

$$\phi_{X_j^2}(t) = \mathbb{E} e^{itX_j^2}$$

$$= \int_{-\infty}^{+\infty} e^{itx^2} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-\mu_j)^2}{2}} dx$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{\mu_j^2}{2}} \int_{-\infty}^{+\infty} e^{(it-\frac{1}{2})x^2 + \mu_j x} dx$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{\mu_j^2}{2}} \int_{-\infty}^{+\infty} e^{(it-\frac{1}{2})\left(x + \frac{\mu_j}{2it-1}\right)^2} e^{-\left(\frac{\mu_j}{2it-1}\right)^2 \cdot (it-\frac{1}{2})} dx$$

$$= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\mu_j^2\left(1 + \frac{1}{2it-1}\right)} \underbrace{\int_{\mathbb{R}} e^{(it-\frac{1}{2})\left(x + \frac{\mu_j}{2it-1}\right)^2} dx}_{\sqrt{\frac{\pi}{-(it-\frac{1}{2})}}}$$

$$= \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{\mu_j^2}{2} \frac{2it}{2it-1}} \cdot \sqrt{\pi} \cdot \left(\frac{1}{2} - it\right)^{-\frac{1}{2}} \sqrt{\frac{\pi}{-(it-\frac{1}{2})}}$$

$$= (1-2it)^{-\frac{1}{2}} \cdot e^{\frac{it\mu_j^2}{1-2it}}$$

since

$$\int_{-\infty}^{+\infty} e^{-ax^2} dx = \sqrt{\frac{\pi}{a}}$$

$$\phi_Y(t) = (1-2it)^{-\frac{n}{2}} \cdot e^{\frac{it}{1-2it} \cdot \sum_{j=1}^n \mu_j^2}$$

for $\forall a \in \mathbb{C}, \operatorname{Re}(a) < 0$.

Remark: $\int_{\mathbb{R}} e^{-ax^2} dx \cdot \int_{\mathbb{R}} e^{-ay^2} dy \stackrel{\text{polar}}{=} \int_0^{2\pi} \int_0^{+\infty} e^{-ar^2} \cdot r \, dr \, d\theta$
 $= \frac{2\pi}{a}$, taking square root on both sides proves the argument (or by analytic continuation)

S.7.10: X, Y cts r.v., show

$$\int_{-\infty}^{+\infty} \phi_X(y) f_Y(y) e^{-ity} \, dy = \int_{-\infty}^{+\infty} \phi_Y(x-t) f_X(x) \, dx$$

pf: RHS = $\int_{-\infty}^{+\infty} \mathbb{E} e^{i(x-t)Y} f_X(x) \, dx$
 $= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} e^{i(x-t)y} f_Y(y) \, dy f_X(x) \, dx$

Fubini $\int_{\mathbb{R}} e^{-ity} f_Y(y) \left(\int_{\mathbb{R}} e^{ixy} f_X(x) \, dx \right) \, dy$
 $= \int_{\mathbb{R}} \phi_X(y) f_Y(y) e^{-ity} \, dy.$
 $\mathbb{E} e^{iyX} = \phi_X(y)$

set $t=0$, $\int_{\mathbb{R}} \phi_X(y) f_Y(y) \, dy = \int_{\mathbb{R}} \phi_Y(x) f_X(x) \, dx$
 $\parallel \mathbb{E} \phi_X(Y) \quad \text{(Parseval relation)} \quad \parallel \mathbb{E} \phi_Y(X)$

e.g. Triangular Dist, $f(x) = 1 - |x|$, $x \in (-1, 1)$

$$\begin{aligned} \text{c.f. } \phi_x(t) &= \int_{-1}^1 (1 - |x|) e^{itx} dx \\ &= \int_0^1 (1 - x) e^{itx} dx + \int_{-1}^0 (1 + x) e^{itx} dx \\ &= \frac{1}{it} \left(-1 + \frac{e^{it} - 1}{it} \right) + \frac{1}{it} \left(1 - \frac{1 - e^{-it}}{it} \right) \\ &= \frac{2 - 2 \cos t}{t^2} \end{aligned}$$

Then for a RW $\{S_n\}$ on \mathbb{R}^d , μ_n as dist of S_n ,

S_n departs from $S_0 = 0$ and want to investigate prob of S_n going back to 0, $\forall \delta > 0$,

$$\mathbb{P}(\|S_n\| < \frac{1}{\delta}) = \int_{\|t\| < \frac{1}{\delta}} d\mu_n(t) \quad \left(\text{if } |x| \leq \frac{\pi}{3}, 1 - \cos x \geq \frac{x^2}{4} \right)$$

$$\left\{ \begin{array}{l} S_n \sim \mu_n \\ X = (X_1, \dots, X_d) \\ X_i \text{ i.i.d. Triangular} \end{array} \right. \leq \int_{\mathbb{R}^d} \prod_{i=1}^d \frac{4}{\delta^2 t_i^2} [1 - \cos(\delta t_i)] d\mu_n(t)$$

$$= 2^d \int_{\mathbb{R}^d} \prod_{i=1}^d \frac{2(1 - \cos(\delta t_i))}{\delta^2 t_i^2} d\mu_n(t)$$

$$\text{apply Parseval relation!} = 2^d \int_{(-\delta, \delta)^d} \prod_{i=1}^d \frac{\delta^{-1|x_i|}}{\delta^2} \phi_{S_n}(x) dx$$

★ Key step in building judging criterion for the recurrence of random walks!

5.8.b: X_1, \dots, X_n be such that $\forall a_1, \dots, a_n \in \mathbb{R}$,

$\sum_{i=1}^n a_i X_i$ is always Gaussian. Show that joint c.f.

of X is $e^{it^T \mu} - \frac{1}{2} t^T V t$ for some $\mu \in \mathbb{R}^n$, $V \in \mathbb{R}^{n \times n}$.

Show that X has multivariate Gaussian density if V is invertible.

Pf:

$$\begin{aligned} \phi_X(t) &= \mathbb{E} e^{it^T X} = \mathbb{E} e^{i(t_1 X_1 + \dots + t_n X_n)} \\ &\quad \text{Gaussian, denote } Y_t \\ &= e^{i \mu(t)} - \frac{1}{2} \sigma^2(t) \end{aligned}$$

for some $\mu, \sigma: \mathbb{R}^n \rightarrow \mathbb{R}$

$$\begin{cases} \mathbb{E} Y_t = \mu(t) = \sum_{i=1}^n t_i \mathbb{E} X_i = t^T \cdot \underbrace{\mu}_{\text{mean vector of } X} \\ \text{Var } Y_t = \sigma^2(t) = t^T \cdot \underbrace{V}_{\text{covariance matrix of } X} t \end{cases}$$

If V is invertible, since V is symmetric and

semi-positive-definite, consider $Z = V^{-\frac{1}{2}}(X - \mu)$

$$\begin{aligned} \phi_Z(t) &= \mathbb{E} e^{it^T V^{-\frac{1}{2}}(X - \mu)} = e^{-it^T V^{-\frac{1}{2}} \mu} \cdot \phi_X(V^{-\frac{1}{2}} t) \\ &= e^{-it^T V^{-\frac{1}{2}} \mu} \cdot e^{it^T V^{-\frac{1}{2}} \mu - \frac{1}{2} t^T V^{-\frac{1}{2}} V V^{-\frac{1}{2}} t} = e^{-\frac{1}{2} t^T t} \end{aligned}$$

So: $\mathbf{z} = \mathbf{V}^{-\frac{1}{2}}(\mathbf{x} - \boldsymbol{\mu}) \sim \mathcal{N}(\mathbf{0}, \mathbf{I}_n)$, which has density as the product of n Gaussian marginals since components are independent.

$$f_{\mathbf{z}}(\mathbf{z}) = (2\pi)^{-\frac{n}{2}} e^{-\frac{\mathbf{z}^T \mathbf{z}}{2}}$$

Then, $\mathbf{X} = \mathbf{V}^{\frac{1}{2}} \mathbf{z} + \boldsymbol{\mu}$ also admits a density

$$f_{\mathbf{X}}(\mathbf{x}) = f_{\mathbf{z}}(\mathbf{V}^{-\frac{1}{2}}(\mathbf{x} - \boldsymbol{\mu})) \cdot \det\left(\frac{\partial \mathbf{z}}{\partial \mathbf{x}}\right) \leftarrow \text{Jacobian matrix}$$

$$= (2\pi)^{-\frac{n}{2}} \cdot e^{-\frac{(\mathbf{x} - \boldsymbol{\mu})^T \mathbf{V}^{-1} (\mathbf{x} - \boldsymbol{\mu})}{2}} \cdot [\det(\mathbf{V})]^{-\frac{1}{2}}$$

c.f. has natural connection with moments:

$$\phi'_X(t) = \mathbb{E}(iX e^{itX}), \text{ so } \phi'_X(0) = i \cdot \mathbb{E}X$$

if $X \in \mathcal{L}^1$

e.g: If $\phi_X(t) = 1 + o(t^2)$ ($t \rightarrow 0$), then $\phi_X \equiv 1$.

Pf: $\lim_{t \rightarrow 0} \frac{\phi_X(t) - 1}{t^2} = 0$, so $\phi'_X(0) = \lim_{t \rightarrow 0} \frac{\phi_X(t) - 1}{t} = 0$

which means $\mathbb{E}X = 0$, and

$$\phi''_X(0) = \lim_{t \rightarrow 0} \frac{\phi_X(t) + \phi_X(-t) - 2\phi_X(0)}{t^2} = 0, \text{ so}$$

$$X \in \mathcal{L}^2 \text{ and } \mathbb{E}X^2 = 0 \Rightarrow \text{Var} X = 0 \Rightarrow X = 0 \text{ a.s.}$$

This implies $\phi_X \equiv 1$, so c.f. cannot be "too flat" in the neighborhood of 0.