

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

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

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  $Y$ .

Bf:

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

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

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

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

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

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

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

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

Since

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

$$\phi_Y(t) = (1-2it)^{-\frac{n}{2}} \cdot e^{\frac{it}{1-2it} \cdot \sum_{j=1}^n \mu_j^2} \quad \text{for } \forall a \in \mathbb{C}, \operatorname{Re} a < 0.$$

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)

5.7.10:  $X, Y$  cts r.v., show

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

$$\begin{aligned} \underline{\text{Qf}}: \text{ RHS} &= \int_{-\infty}^{+\infty} \mathbb{E} e^{i(x-t)Y} f_X(x) dx \\ &= \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} e^{i(x-t)y} f_Y(y) dy f_X(x) dx \end{aligned}$$

$$\begin{aligned} \underline{\text{Fab. ni}}: & \int_{\mathbb{R}} e^{-ity} f_Y(y) \underbrace{\int_{\mathbb{R}} e^{ixy} f_X(x) dx dy}_{\mathbb{E} e^{iyX} = \phi_X(y)} \\ &= \int_{\mathbb{R}} \phi_X(y) f_Y(y) e^{-ity} dy. \end{aligned}$$

$$\underline{\text{set } t=0,} \quad \int_{\mathbb{R}} \phi_X(y) f_Y(y) dy = \int_{\mathbb{R}} \phi_Y(x) f_X(x) dx$$

$\mathbb{E} \phi_X(Y) \quad (\text{Parseval relation}) \quad \mathbb{E} \phi_Y(X)$

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

$$\text{c.f. } \phi_x(t) = \int_{-1}^1 (1 - |x|) e^{itx} dx$$

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

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

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

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

$$\left\{ S_n \sim \mu_n \right. \leq \int_{\mathbb{R}^d} \prod_{i=1}^d \frac{4}{\delta^2 t_i^2} [1 - \cos(\delta t_i)] d\mu_n(t)$$

$$\left\{ \begin{array}{l} X = (X_1, \dots, X_d) \\ X_i \stackrel{i.i.d.}{\sim} \text{Triangular} \end{array} \right. = 2^d \int_{\mathbb{R}^d} \prod_{i=1}^d \frac{2(1 - \cos(\delta t_i))}{\delta^2 t_i^2} d\mu_n(t) \quad \text{c.f. dist}$$

$$\left. \begin{array}{l} \text{apply Parseval relation!} \\ = 2^d \int_{(-\delta, \delta)^d} \prod_{i=1}^d \frac{\delta - |x_i|}{\delta^2} \phi_{S_n}(x) dx \end{array} \right. \quad \text{c.f. density.}$$

★ 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:  $\phi_X(t) = \mathbb{E} e^{it^T X} = \mathbb{E} e^{i(t_1 X_1 + \dots + t_n X_n)}$   
Gaussian, denote  $Y_t$

$$= e^{i \cdot \mu(t) - \frac{1}{2} G^2(t)}$$

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

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

If  $V$  is invertible, since  $V$  is symmetric and semi-positive-definite, consider  $Z = V^{-\frac{1}{2}}(X - \mu)$

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

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

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

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

$$f_X(x) = f_{\tilde{z}}(V^{-\frac{1}{2}}(x - \mu)) \cdot \det\left(\frac{\partial \tilde{z}}{\partial x}\right) \xrightarrow{\text{Jacobian matrix}}$$

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

c.f. has natural connection with moments:

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

if  $X \in \mathbb{Z}^1$

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

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

$X \in \mathbb{Z}^2$  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.