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Exercise 27. First, we want show that

$$m_k := \mathbb{E}[X^k] = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} x^k e^{-\frac{x^2}{2}} dx = \begin{cases} (k-1)!! & \text{if } k \text{ is even;} \\ 0 & \text{if } k \text{ is odd.} \end{cases}$$

Let's define

$$I(k) := \int_{\mathbb{R}} x^k e^{-\frac{x^2}{2}} dx.$$

If we can prove the following claim, we are done.

<u>Claim:</u> For all  $p \in \mathbb{N}$ , we have

1. 
$$I(2p) = (2p-1)!!\sqrt{2\pi}$$
;

2. 
$$I(2p+1) = 0$$
.

Proof: Both the proofs proceed by induction on p.

1. Suppose p = 0. Then

$$I(2p) = I(0) = \int_{\mathbb{D}} e^{-\frac{x^2}{2}} dx = \sqrt{2\pi} = (2 \cdot 0 - 1)!! \sqrt{2\pi}$$

Suppose now p > 0. We have

$$\begin{split} I(2p) &= \int_{\mathbb{R}} x^{2p} e^{-\frac{x^2}{2}} dx \\ &= \int_{\mathbb{R}} -x^{2p-1} \left[ -x e^{-\frac{x^2}{2}} \right] dx \\ &= -x^{2p-1} e^{-\frac{x^2}{2}} \bigg|_{-\infty}^{+\infty} + (2p-1) \int_{\mathbb{R}} x^{2p-2} e^{-\frac{x^2}{2}} dx \\ &= 0 + (2p-1)[(2p-3)!! \sqrt{2\pi}] \\ &= (2p-1)!! \sqrt{2\pi}. \end{split}$$

2. For p = 0 we obtain

$$I(2p+1) = I(1) = \int_{\mathbb{R}} xe^{-\frac{x^2}{2}} dx = \mathbb{E}[X] = 0.$$

For p > 0, proceeding similarly to point (1), we obtain

$$I(2p+1) = \int_{\mathbb{R}} x^{2p+1} e^{-\frac{x^2}{2}} dx = -x^{2p} e^{-\frac{x^2}{2}} \bigg|_{-\infty}^{+\infty} + 2p \int_{\mathbb{R}} x^{2p-1} e^{-\frac{x^2}{2}} dx = 0 + 0 = 0.$$

Now we want to show that

$$|m|_k := \mathbb{E}[|X|^k] = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} |x|^k e^{-\frac{x^2}{2}} dx = \begin{cases} (k-1)!! & \text{if } k \text{ is even;} \\ (k-1)!! \sqrt{\frac{2}{\pi}} & \text{if } k \text{ is odd.} \end{cases}$$

If k is even, the statement follows trivially by  $m_k = |m|_k$ . Suppose then p odd. If we can prove the following claim, we are done.

Claim: For all  $p \in \mathbb{N}$ , we have

$$\int_{\mathbb{R}} |x|^{2p+1} e^{-\frac{x^2}{2}} dx = (2p)!! \cdot 2.$$

<u>Proof:</u> The proof proceeds by induction on p. Suppose p = 0. Then

$$\int_{\mathbb{R}} |x|^{2p+1} e^{-\frac{x^2}{2}} dx = \int_{\mathbb{R}} |x| e^{-\frac{x^2}{2}} dx = 2 \int_0^{+\infty} |x| e^{-\frac{x^2}{2}} dx = -2e^{-\frac{x^2}{2}} \bigg|_0^{+\infty} = 0 + 2 = (2 \cdot 0)!! \cdot 2.$$

Suppose now p > 0. We have

$$I(2p+1) = \int_{\mathbb{R}} |x|^{2p+1} e^{-\frac{x^2}{2}} dx$$

$$= \int_{\mathbb{R}} -x^{2p} \left[ -|x| e^{-\frac{x^2}{2}} \right] dx$$

$$= 2 \int_0^{+\infty} -x^{2p} \left[ -x e^{-\frac{x^2}{2}} \right] dx$$

$$= -2x^{2p} e^{-\frac{x^2}{2}} \Big|_0^{+\infty} + 4p \int_0^{+\infty} x^{2p-1} e^{-\frac{x^2}{2}} dx$$

$$= 0 + 2p \int_{\mathbb{R}} |x|^{2p-1} e^{-\frac{x^2}{2}} dx$$

$$= 2p[(2p-2)!! \cdot 2] = (2p)!! \cdot 2$$

In order to find the characteristic function, we will soon need the following:

**Lemma 0.0.1.** For every  $p \in \mathbb{N}_{\geq 1}$  we have

$$\log 1 \log 2 + \dots + \log p \ge \int_1^p \log x dx.$$

Proof.

$$\log 1 + \log 2 + \dots + \log p = 0 + \log 2 \int_{1}^{2} 1 dx + \log 3 \int_{2}^{3} 1 dx + \dots + \log p \int_{p-1}^{p} 1 dx$$

$$= \int_{1}^{2} \log 2 dx + \int_{2}^{3} \log 3 dx + \dots + \int_{p-1}^{p} \log p dx$$

$$\geq \int_{1}^{2} \log x dx + \int_{2}^{3} \log x dx + \dots + \int_{p-1}^{p} \log x dx$$

$$= \int_{1}^{p} \log x dx,$$

where the inequality holds since log is an increasing function, thus  $\log x \leq \log(i+1)$  for all  $x \in [i, i+1]$ .

To lighten the notation, suppose WLOG that k=2p for some  $p\in\mathbb{N}.$  Observe now that

$$\lim_{k \to +\infty} \left( \frac{|m|_k}{k!} \right)^{\frac{1}{k}} \leq \lim_{k \to +\infty} \left( \frac{(k-1)!!}{k!} \right)^{\frac{1}{k}}$$

$$= \lim_{k \to +\infty} \left( \frac{1}{k!!} \right)^{\frac{1}{k}} \qquad \text{(since } k!! \geq (k/2)!)$$

$$\leq \lim_{k \to +\infty} \left( \frac{1}{(k/2)!} \right)^{\frac{1}{k}}$$

$$= \lim_{p \to +\infty} \left( \frac{1}{p!} \right)^{\frac{2}{p}}$$

$$= \lim_{p \to +\infty} e^{\frac{1}{p} \log \left( \frac{1}{p!} \right)}$$

$$= \lim_{p \to +\infty} e^{-\frac{2}{p} (\log p!)}$$

$$= \lim_{p \to +\infty} e^{-\frac{2}{p} (\log 1 + \log 2 + \dots + \log p)}$$

$$\leq \lim_{p \to +\infty} e^{-\frac{2}{p} \int_1^p \log x dx} \qquad \text{(thanks to the lemma)}$$

$$= \lim_{p \to +\infty} e^{-\frac{2}{p} (p \log p - p + 1)}$$

$$= \lim_{p \to +\infty} e^{-2 \log p + 2 - \frac{2}{p}} = e^{-\infty} = 0.$$

This means that we can apply the corollary we saw during the lecture, i.e.

$$\varphi(t) = \sum_{k=0}^{\infty} m_k \frac{(it)^k}{k!}.$$

Since  $m_k = 0$  if k is odd, we obtain

$$\sum_{k=0}^{\infty} m_k \frac{(it)^k}{k!} = \sum_{h=0}^{\infty} m_{2h} \frac{(it)^{2h}}{(2h)!} = \sum_{h=0}^{\infty} (2h-1)!! (-1)^h \frac{t^{2h}}{(2h)!} = \sum_{h=0}^{\infty} (-1)^h \frac{t^{2h}}{(2h)!!}.$$

It is very easy to check that  $(2h)!! = 2^h h!$ , for all  $h \in \mathbb{N}$ . Therefore

$$\sum_{h=0}^{\infty} (-1)^h \frac{t^{2h}}{(2h)!!} = \sum_{h=0}^{\infty} (-1)^h \frac{t^{2h}}{2^h h!} = \sum_{h=0}^{\infty} \frac{\left(-\frac{t^2}{2}\right)^h}{h!} = e^{-\frac{t^2}{2}}.$$

That is,  $\varphi(t) = e^{-\frac{t^2}{2}}$ .

## Exercise 28.

$$\varphi_{aX+b}(t) = \mathbb{E}\left[e^{it(aX+b)}\right] = \mathbb{E}\left[e^{itaX}e^{itb}\right] = \mathbb{E}\left[e^{itaX}\right]\mathbb{E}\left[e^{itb}\right] = \varphi_X(at)e^{itb}.$$

Thus, since  $X \sim N(\mu, \sigma^2) \Rightarrow X = \sigma Y + \mu$  with  $Y \sim N(0, 1)$ , we have

$$\varphi_X(t) = \varphi_{\sigma Y + \mu}(t) = e^{-\frac{\sigma^2 t^2}{2}} e^{it\mu} = e^{it\mu - \frac{1}{2}\sigma^2 t^2}.$$

## Exercise 29. We easily have

$$\varphi_{-X}(t) = \mathbb{E}\left[e^{i(-t)X}\right] = \varphi_X(-t)$$

and thus

$$\varphi_{X-Y}(t) = \mathbb{E}\left[e^{it(X+(-Y))}\right] = \varphi_X(t)\varphi_X(-t).$$

Now:

a) Consider  $f(t) := e^{2(\cos t - 1)}$ . We want to find explicity a random variable Z such that  $\varphi_Z(t) = f(t)$ . Let Y be a Bernoulli r.v. with  $\mathbb{P}[0] = \mathbb{P}[1] = \frac{1}{2}$ . Then

$$\varphi_Y(t) = \frac{1}{2}e^{-it} + \frac{1}{2}e^{it} = \cos t.$$

Now let N be distributed like a Poisson distribution of parameter  $\lambda$ , and let  $(Y_i)$  be i.i.d. random variables, independent with N. Define

$$S = \sum_{j=1}^{N} Y_j.$$

We have

$$\varphi_{S}(t) = \mathbb{E}\left[e^{itS}\right] = \mathbb{E}\left[e^{it\sum_{j=1}^{N}Y_{j}}\right]$$

$$= \mathbb{E}\left[\sum_{n\geq0}e^{it\sum_{j=1}^{n}Y_{j}}\mathbb{1}_{\{N=n\}}\right]$$

$$= \sum_{n\geq0}\mathbb{E}\left[e^{it\sum_{j=1}^{n}Y_{j}}\right] \cdot \mathbb{E}\left[\mathbb{1}_{\{N=n\}}\right] \qquad (N \text{ and } (Y_{j}) \text{ indep.})$$

$$= \sum_{n\geq0}\mathbb{E}\left[e^{it\sum_{j=1}^{n}Y_{j}}\right] \cdot \mathbb{P}[N=n]$$

$$= \sum_{n\geq0}\underbrace{\varphi_{Y_{1}}(t)^{n}}_{\mathbb{E}e^{it\sum_{j=1}^{n}Y_{j}}} \cdot e^{-\lambda}\frac{\lambda^{n}}{n!} = e^{\lambda(\varphi_{Y_{1}}(t)-1)}.$$

Therefore, if we take

$$Z := \sum_{j=1}^{N} Y_j$$

with  $(Y_j)$  i.i.d. Bernoulli symmetric r.v. with range  $\{1, -1\}$  and  $N \sim \text{Pois}(2)$  independent with  $(Y_j)$ , we obtain

$$\varphi_Z(t) = e^{2(\cos t - 1)},$$

as wanted.

• Consider  $f(t) = e^{-|t|^3}$ . We have

$$f'(t) = -3|t|te^{-|t|^3}$$

and

$$f''(t) = -3\left(|t|e^{-|t|^3} + |t|e^{-|t|^3} - 3|t|te^{-|t|^3}|t|t\right).$$

So f'(0) = f''(0) = 0. Let X be a random variable. We now that  $\varphi_X'(0) = i \mathbb{E}[X]$ . So, if  $f(t) = \varphi_X(t)$ , we would have  $\mathbb{E}[X] = 0$ . But since  $\varphi_X''(0) = -\mathbb{E}[X^2] = \mathbb{E}[(X-0)^2] = \mathrm{Var}[X]$ , we would have  $\mathrm{Var}[X] = 0$ . Therefore X must be the constant function in 0. But the characteristic function of  $X \equiv 0$  is trivially  $e^{it0} = 1 \neq e^{-|t|^3}$ . Thus  $e^{-|t|^3}$  is not the characteristic function of any random variable.