

# STAT 430-1, Fall 2024

## HW5

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**Notation:** I use  $\hat{=}$  for Fourier transform and  $\hat{=}$  for inverse Fourier transform. i.e.  $f(x) \hat{=} \phi(t)$  and  $\phi(t) \hat{=} f(x)$ .  $\text{Res}_f(x)$  for the residue of  $f(x)$  at  $x$ . Dirac delta function (at zero) is denoted as  $\delta(x)$  s.t.  $\int_{-\varepsilon}^{\varepsilon} \delta(x) dx = 1, \forall \varepsilon > 0$ .

### Exercise 1 Characteristic functions

(a) Note that function  $t \mapsto 1/(1+t^2)$  has inverse Fourier transform as follows:

$$1/(1+t^2) \hat{=} \frac{1}{2\pi} \int_{\mathbb{R}} \frac{1}{1+t^2} e^{-ixt} dt = \begin{cases} \frac{2\pi i}{2\pi} \text{Res}_{e^{-ixt}/(1+t^2)}(i), & x > 0 \\ \frac{2\pi i}{2\pi} \text{Res}_{e^{-ixt}/(1+t^2)}(-i), & x \leq 0 \end{cases} = e^{-|x|}/2$$

which is the density function of a two-sided exponential distribution.

(b) Not a characteristic function by noticing that the characteristic function  $t \mapsto e^{-t^4} := \phi_X(t)$  yields a 0 second moment:

$$\mathbb{E}[X^2] = \phi_X''(0) = \frac{d^2}{dt^2} e^{-t^4} \Big|_{t=0} = 0$$

which means that  $X$  is actually a degenerate r.v. with  $X = 0$  a.s., which should actually correspond to a characteristic function of  $t \mapsto 1$  instead. Thus the contradiction arises, so  $t \mapsto e^{-t^4}$  is not a characteristic function.

**Reference:** I checked the proof in this [Mathexchange post](#).

(c) Not a characteristic function by noticing that  $\sin(t)|_{t=0} = 0 \neq 1$ , thus does not satisfy the condition of Bochner's theorem.

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(d) Note that  $t \mapsto \cos(t)$  has inverse Fourier transform as follows:

$$\cos(t) \doteq \frac{1}{2\pi} \int_{\mathbb{R}} \cos(t) e^{-ixt} dt = \frac{1}{2\pi} \int_{\mathbb{R}} \frac{e^{it} + e^{-it}}{2} e^{-ixt} dt = \frac{1}{2} (\delta(x+1) + \delta(x-1)) \sim \text{Unif}\{\pm 1\}$$

a uniform distribution on  $\{\pm 1\}$ .

(e) Note that  $t \mapsto \frac{1 + \cos(t)}{2}$  has inverse Fourier transform as follows:

$$\frac{1 + \cos(t)}{2} \doteq \frac{1}{2\pi} \int_{\mathbb{R}} \frac{1 + \cos(t)}{2} e^{-ixt} dt = \frac{1}{2\pi} \int_{\mathbb{R}} \frac{1 + \frac{e^{it} + e^{-it}}{2}}{2} e^{-ixt} dt = \frac{1}{4} (\delta(x-1) + \delta(x+1) + 2\delta(x))$$

which is a discrete r.v. with  $P(X=1) = P(X=-1) = 1/4$  and  $P(X=0) = 1/2$ .

## Exercise 2 Die roll plus uniform

(a) For a fair die  $X$ , we have

$$X \sim \text{Unif}\{1, 2, 3, 4, 5, 6\} \doteq \frac{1}{6} \sum_{k=1}^6 e^{ikt}$$

(b) For  $Y$  uniform on  $[0, 1]$ , we have

$$Y \sim \text{Unif}[0, 1] \doteq \frac{e^{it} - 1}{it}$$

(c) For  $Z = X + Y$ , we have

$$X + Y \doteq \frac{1}{6} \sum_{k=1}^6 e^{ikt} \cdot \frac{e^{it} - 1}{it} = \frac{1}{6} \sum_{k=1}^6 \frac{e^{i(k+1)t} - e^{ikt}}{it} = \frac{e^{i7t} - e^{it}}{6it} \doteq \text{Unif}[1, 7]$$

(d) For  $W = X - Y$ , we have

$$X - Y \doteq \frac{1}{6} \sum_{k=1}^6 e^{ikt} \cdot \frac{\overline{e^{it} - 1}}{it} = \frac{1}{6} \sum_{k=1}^6 e^{ikt} \cdot \frac{e^{-it} - 1}{-it} = \frac{1}{6} \sum_{k=1}^6 \frac{e^{i(k-1)t} - e^{ikt}}{-it} = \frac{e^{i6t} - 1}{6it} \doteq \text{Unif}[0, 6]$$

## Exercise 3 Difference of two i.i.d. random variables

Note that for two i.i.d. r.v.  $X, Y$  with characteristic function  $\phi(t)$ , we have

$$X - Y \doteq \phi(t) \cdot \overline{\phi(t)} = |\phi(t)|^2 \geq 0$$

however for uniform distribution  $Z \sim \text{Unif}[-1, 1]$  we know that

$$Z \sim \text{Unif}[-1, 1] \doteq \frac{\sin t}{t} \text{ is not non-negative}$$

which yields contradiction. Thus  $Z$  cannot be represented as the difference of two i.i.d. r.v.s.

### Exercise 4 Trigonometric identities

We define the following:

$$X_n \sim \text{Unif}\{\pm \frac{1}{2^n}\} \doteq \frac{1}{2}(e^{i/2^n t} + e^{-i/2^n t}) = \cos(\frac{t}{2^n})$$

$$Y_n := \sum_{i=1}^n X_i \sim \text{Unif}\{\frac{k}{2^n}\}_{k=-2^{n-1}}^{2^n-1}$$

$$Z \sim \text{Unif}[-1, 1] \doteq \frac{\sin t}{t}$$

- We note that  $Z \stackrel{d}{=} \text{Unif}(-\frac{1}{2}, \frac{1}{2}) + X_1$  (which is easy to verify following the same proof as in Exercise. 2). We thus have

$$\begin{aligned} \frac{\sin t}{t} &\doteq Z \stackrel{d}{=} \text{Unif}(-\frac{1}{2}, \frac{1}{2}) + X_1 \doteq \frac{\sin(t/2)}{t/2} \cdot \cos(t/2) \\ \Rightarrow \frac{\sin t}{t} &= \frac{\sin(t/2)}{t/2} \cdot \cos(t/2) \end{aligned}$$

- We further have  $Y_n \xrightarrow{d} Z$  by noticing the following:

$$\forall \xi \in [-1, 1] : \mathbb{P}(Y_n \geq \xi) \xrightarrow{n \rightarrow \infty} \mathbb{P}(Z \geq \xi).$$

Thus we can conclude convergence in characteristic functions:

$$\begin{aligned} \prod_{i=1}^n \cos(\frac{t}{2^i}) &\doteq \sum_{i=1}^n X_i = Y_n \xrightarrow{d} Z \doteq \frac{\sin t}{t} \\ \Rightarrow \frac{\sin t}{t} &= \prod_{i=1}^{\infty} \cos(\frac{t}{2^i}), \quad \forall t \in \mathbb{R}. \end{aligned}$$

### Exercise 5 Poisson approximation to the binomial; exponential approximation to the geometric

(This time we prove them "by hand")

#### 5.(a) Poisson approximation to the Binomial distribution

We have  $\forall k \in \mathbb{N}^+$  :

$$\begin{aligned} \mathbb{P}(\text{Bin}(n, p_n) = k) &= \binom{n}{k} p_n^k (1 - p_n)^{n-k} \\ &= \frac{1}{k!} \frac{n!}{(n-k)!} \cdot \frac{1}{n^n} (np_n)^k (n - np_n)^{n-k} \\ &= \frac{1}{k!} (n^k + o(n^k)) \cdot \frac{1}{n^n} (\lambda + o(1))^k (n - \lambda + o(1))^{n-k} \\ &= \frac{(\lambda + o(1))^k}{k!} \frac{(n^k + o(n^k)) \cdot (n - \lambda + o(1))^{n-k}}{n^n} \\ &= \frac{(\lambda + o(1))^k}{k!} \left(1 - \frac{\lambda}{n} + o(n^{-1})\right)^n \\ &\xrightarrow{n \rightarrow \infty} \frac{\lambda^k}{k!} e^{-\lambda} \sim \text{Poi}(\lambda) \end{aligned}$$

Thus we have proved the Poisson approximation to the Binomial distribution.

## 5.(b) Exponential approximation to the Geometric distribution

We have  $\forall t \geq 0$  that

$$\mathbb{P}(pX_p > t) = \mathbb{P}(\text{Geom}(p) > t/p) = \sum_{k=\lceil t/p \rceil}^{\infty} (1-p)^{k-1}p = (1-p)^{\lfloor t/p \rfloor}$$

Note that

$$(1-p)^{t/p} \xrightarrow{p \rightarrow 0} e^{-t}, \quad (1-p)^{t/p-1} \xrightarrow{p \rightarrow 0} e^{-t}, \quad (1-p)^{t/p-1} \leq (1-p)^{\lfloor t/p \rfloor} \leq (1-p)^{t/p}$$

we have

$$\mathbb{P}(pX_p > t) = (1-p)^{\lfloor t/p \rfloor} \xrightarrow{p \rightarrow 0} e^{-t} = \mathbb{P}(\text{Exp}(1) > t)$$

Thus we have proved the Exponential approximation to the Geometric distribution.