# MATH 418/544 Assignment 6

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## Problem 1(a).

*Proof.* By definition, the characteristic function of geometric distribution with  $p \in (0,1)$  is

$$\phi_X(t) = E[e^{itx}] = \sum_{n=1}^{\infty} e^{itn} P_X(n) = \sum_{n=1}^{\infty} e^{itn} P(X=n) = \sum_{n=1}^{\infty} e^{itn} p(1-p)^{n-1}$$

$$= pe^{it} \sum_{n=1}^{\infty} \left[ e^{it} (1-p) \right]^{n-1} = \frac{pe^{it}}{1 - (1-p)e^{it}} = \frac{p}{e^{-it} - (1-p)}$$
(1)

## Problem 1(b).

*Proof.* The density of the unifrom distribution is:  $f_X(x) = \frac{1}{2a} 1_{[-a,a]}(x)$ . Therefore by definition, the characteristic function of uniform distribution is:

$$\phi_X(t) = E[e^{itx}] = \int_{-a}^{a} \frac{e^{itx}}{2a} dx = \int_{-a}^{a} \left[ \frac{\cos(tx)}{2a} + i \frac{\sin(tx)}{2a} \right] dx = \int_{-a}^{a} \frac{\cos(tx)}{2a} dx$$
 (2)

When t=0, we have  $\phi_X(t)=\int_{-a}^a\frac{1}{2a}dx=1$  and when  $t\neq 0$ , we have  $\phi_X(t)=\frac{\sin(tx)}{2at}|_{-a}^a=\frac{\sin(at)}{at}$ . In summary, the characteristic function is:

$$\phi_X(t) = \begin{cases} \frac{\sin(at)}{at} & \text{if } t \neq 0\\ 1 & \text{if } t = 0 \end{cases}$$

#### Problem 2.

*Proof.* From the question we have

 $p_{k} = (2\pi)^{-1} \int_{-\pi}^{\pi} e^{-itk} \phi_{X}(t) dt = (2\pi)^{-1} \int_{-\pi}^{\pi} \int e^{-itk} e^{itx} dP dt$   $= (2\pi)^{-1} \int_{-\pi}^{\pi} e^{it(x-k)} dt dP \text{ by Fubini's since } |e^{it(x-k)}| = 1 \text{ is integrable}$ (3)

Notice that X is integer valued and  $k \in \mathbb{Z}$ , so  $x - k \in \mathbb{Z}$ . Then we have  $\int_{-\pi}^{\pi} \frac{e^{it(x-k)}}{2\pi} = 1_{(x=k)}$ , and hence

$$p_k = (2\pi)^{-1} \int 1_{(x=k)} dP = P(x=k)$$
(4)

which is the the probability of mass function at k by definition.

### Problem 3.

*Proof.* Let  $X_i$  be the round-off of the ith number, then  $\{X_n:n\in\mathbb{N},n\leq 25\}$  are a sequence of i.i.d random variables. Then we have  $E(X_i)=0,\ \sigma^2=E(X_i^2)-E(X_i)^2=\int_{-0.5}^{0.5}x^2dx=\frac{x^3}{3}|_{-0.5}^{0.5}=\frac{1}{12}<\infty.$  Denote

 $S_n = \sum_{i=1}^n X_i$ , since the sum of rounded numbers equals to rounded sum of the unrounded numbers, we have:

$$P(-0.5 < S_{25} < 0.5) = P(\frac{-0.5}{\sqrt{25/12}} < \frac{S_{25}}{\sqrt{\sigma^2 25}} < \frac{0.5}{\sqrt{25/12}})$$

$$\approx P(-\frac{\sqrt{3}}{5} < \frac{S_{25}}{\sqrt{\sigma^2 25}} < \frac{\sqrt{3}}{5})$$
where  $\frac{S_{25}}{\sqrt{\sigma^2 25}} \sim N(0, 1)$  by Central Limit Theorem
$$\approx P(-0.35 < Z < 0.35) \text{ where } Z = \frac{S_{25}}{\sqrt{\sigma^2 25}}$$

$$\approx 0.2736$$
(5)

Problem 4.

Proof. Let  $\{X_i: i \in \mathbb{N}\}$ ,  $\{Y_i: i \in \mathbb{N}\}$  be sequences of i.i.d random variables, where  $X_i, Y_i$  is the time to make the *i*th double decaf skim milk for George and Julia respectively, and  $\{X_i\}$ ,  $\{Y_i\}$  are independent. Then we have  $E[X_i] = E[Y_i]$  and same standard deviation  $\sigma_{X_i} = \sigma_{Y_i} = 4$ . Denote  $D_i = Y_i - X_i$ , then we have  $E[D_i] = E[Y_i] - E[X_i] = 0$  and  $\sigma_{D_i}^2 = \sigma_{Y_i}^2 + \sigma_{X_i}^2 = 32$ . Also denote  $S_n = \sum_{i=1}^n D_i$ , then the probability that Goerge will get the prize is:

$$P(S_{200} > 80) = P(\sum_{i=1}^{n} D_{i} > 80) = P(\frac{S_{200}}{\sqrt{200\sigma_{D_{i}}^{2}}} \ge \frac{80}{\sqrt{200\sigma_{D_{i}}^{2}}}) = P(\frac{S_{200}}{\sqrt{200\sigma_{D_{i}}^{2}}} \ge \frac{80}{\sqrt{200 \times 32}})$$

$$\approx P(Z \ge 1) \text{ by Central limit theorem, where } Z = \frac{S_{200}}{\sqrt{200\sigma_{D_{i}}^{2}}} \sim N(0, 1)$$

$$= \int_{1}^{\infty} \frac{e^{-x^{2}/2}}{\sqrt{2\pi}} dx$$
(6)

Problem 5.

*Proof.* Since  $\{X_n\}$  are i.i.d Cauchy random variables and  $S_n = \sum_{k=1}^n X_k$ ,

$$\lim_{n \to \infty} \phi_{S_n/n}(t) = \lim_{n \to \infty} E[e^{it(S_n/n)}] = \lim_{n \to \infty} E[e^{i\frac{t}{n}S_n}]$$

$$= \lim_{n \to \infty} \phi_{S_n}(\frac{t}{n}) = [\phi_{X_1}(\frac{t}{n})]^n = (e^{-|t|/n})^n = e^{-|t|} = \phi_{X_1}(t)$$
(7)

Since  $\phi_{S_n/n}(t)$  converges for all t and the limit  $\phi_{X_1}(t)$  is continuous at 0, by Levy's continuity theorem we know  $S_n/n$  has the same distribution as  $X_1$ , and hence the limiting distribution is Cauchy.