

Math 733 - Fall 2020

Homework 3

Due: 10/11, 10pm

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1. (a) *Proof.*

$$X \sim B(n, p) \Rightarrow P(X = k) = \binom{n}{k} p^k (1-p)^{n-k}$$

$$Y \sim B(m, p) \Rightarrow P(Y = k) = \binom{m}{k} p^k (1-p)^{m-k}$$

Then

$$\begin{aligned} P(X + Y = k) &= \sum_{i=0}^k P(X = i, Y = k - i) \\ &= \sum_{i=0}^k P(X = i) \cdot P(Y = k - i) \\ &= \sum_{i=0}^k \binom{n}{i} p^i (1-p)^{n-i} \cdot \binom{m}{k-i} p^{k-i} (1-p)^{m-k+i} \\ &= p^k (1-p)^{m+n-k} \sum_{i=0}^k \binom{n}{i} \binom{m}{k-i} \\ &= \binom{n+m}{k} p^k (1-p)^{m+n-k} \end{aligned}$$

Thus,

$$X + Y \sim B(n + m, p)$$

□

(b) *Proof.*

$$X \sim \text{Poisson}(\lambda) \Rightarrow P(X = k) = \frac{\lambda^k}{k!} e^{-\lambda}$$

$$Y \sim \text{Poisson}(\mu) \Rightarrow P(Y = k) = \frac{\mu^k}{k!} e^{-\mu}$$

Then

$$\begin{aligned}
P(X + Y = k) &= \sum_{i=0}^k P(X = i, Y = k - i) \\
&= \sum_{i=0}^k P(X = i) \cdot P(Y = k - i) \\
&= \sum_{i=0}^k \frac{\lambda^i}{i!} e^{-\lambda} \cdot \frac{\mu^{k-i}}{(k-i)!} e^{-\mu} \\
&= e^{-(\lambda+\mu)} \sum_{i=0}^k \frac{\lambda^i}{i!} \frac{\mu^{k-i}}{(k-i)!} \\
&= \frac{(\lambda + \mu)^k}{k!} e^{-(\lambda+\mu)}
\end{aligned}$$

Thus,

$$X + Y \sim \text{Poisson}(\lambda + \mu)$$

□

2. (a) *Proof.* Let $h(x, y) = \mathbb{1}_{\{xy \leq z\}}$, let μ, ν be the probability measures with distributions F_X and F_Y . Since for fixed $y > 0$,

$$\int h(x, y) \mu(dx) = \int \mathbb{1}_{(-\infty, z/y]}(x) \mu(dx) = F_X\left(\frac{z}{y}\right)$$

So

$$\begin{aligned}
F_{XY}(z) &= P(XY \leq z) = \iint \mathbb{1}_{\{xy \leq z\}} \mu(dx) \nu(dy) \\
&= \int F_X\left(\frac{z}{y}\right) dF_Y(y)
\end{aligned}$$

□

- (b) *Proof.* Absolutely continuous means every set of measure zeros is probability zero. Consider

$$P(XY) = P(X)P(Y)$$

$$P(XY = x_1 y_1) = P(X = x_1)P(Y = y_1) = 0$$

Then XY is absolutely continuous with p.d.f

$$f_{XY} = \int f_X\left(\frac{x}{t}\right) f_Y(t) dt$$

□

3. *Proof.*

$$\begin{aligned}
&\lim_{n \rightarrow \infty} P(|X_n + Y_n - (X + Y)| > \varepsilon) \\
&\leq \lim_{n \rightarrow \infty} P\left(|X_n - X| > \frac{\varepsilon}{2}\right) + \lim_{n \rightarrow \infty} P\left(|Y_n - Y| > \frac{\varepsilon}{2}\right) \\
&= 0
\end{aligned}$$

Since $X_n \xrightarrow{p} X$ and $Y_n \xrightarrow{p} Y$. So,

$$X_n + Y_n \xrightarrow{p} X + Y$$

$$\begin{aligned}
& \lim_{n \rightarrow \infty} P(|X_n Y_n - XY| > \varepsilon) \\
&= \lim_{n \rightarrow \infty} P(|(X_n - X)(Y_n - Y) + Y(X_n - X) + X(Y_n - Y)| > \varepsilon) \\
&\leq \lim_{n \rightarrow \infty} P\left(|(X_n - X)(Y_n - Y)| > \frac{\varepsilon}{3}\right) + \lim_{n \rightarrow \infty} P\left(|Y(X_n - X)| > \frac{\varepsilon}{3}\right) + \lim_{n \rightarrow \infty} P\left(|X(Y_n - Y)| > \frac{\varepsilon}{3}\right) \\
&= 0
\end{aligned}$$

$$X_n Y_n \xrightarrow{P} XY$$

□

4. *Proof.* Consider

$$x_i \sim U[0, 1] \forall i \text{ and i.i.d}$$

$$\begin{aligned}
& \lim_{n \rightarrow \infty} \int_0^1 \int_0^1 \cdots \int_0^1 n \left(f\left(\frac{1}{n}(x_1 + x_2 + \cdots + x_n)\right) - f\left(\frac{1}{2}\right) \right) dx_1 dx_2 \cdots dx_n \\
&= \lim_{n \rightarrow \infty} \mathbb{E} \left[n \left(f\left(\frac{1}{n}(x_1 + x_2 + \cdots + x_n)\right) - f\left(\frac{1}{2}\right) \right) \right] \\
&= \lim_{n \rightarrow \infty} n \cdot \mathbb{E} \left[f\left(\frac{1}{n}(x_1 + x_2 + \cdots + x_n)\right) - f\left(\frac{1}{2}\right) \right] \\
& f \text{ is continuous differentiable} \Leftrightarrow \mathbb{E} \left[f\left(\frac{x_1 + x_2 + \cdots + x_n}{n}\right) \right] = f\left(\mathbb{E} \left[\frac{x_1 + x_2 + \cdots + x_n}{n} \right]\right) \\
&= \lim_{n \rightarrow \infty} n \cdot \left(f\left(\mathbb{E} \left[\frac{x_1 + x_2 + \cdots + x_n}{n} \right]\right) - f\left(\frac{1}{2}\right) \right) \\
& \text{Since } \mathbb{E} \left[\frac{x_1 + x_2 + \cdots + x_n}{n} \right] = \mathbb{E}[x_1] = \int_0^1 t dt = \frac{1}{2} \\
& \text{We have} \\
&= \lim_{n \rightarrow \infty} n \cdot 0 = 0
\end{aligned}$$

□

5. (a) *Proof.* Let

$$S_n = x_1 + x_2 + \cdots + x_n$$

then

$$\mathbb{E}[\mathbb{1}_{\{S_k \leq x\}}] = P(S_k \leq x) = P(x_1 + x_2 + \cdots + x_k < k) = F^{(k)}(x)$$

Let

$$N_x = \sum_{k=1}^{\infty} \mathbb{1}_{\{S_k \leq x\}}$$

It gives

$$\mathbb{E}[N_x] = \sum_{n=1}^{\infty} \mathbb{E}[\mathbb{1}_{\{S_n \leq x\}}] = \sum_{n=1}^{\infty} F^{(n)}(x)$$

□

(b) *Proof.* Let $n = \lceil x \rceil$

$$\begin{aligned}
P(N_x = N) &= P(S_N \leq x < S_{N+1}) \\
&\leq P(S_N \leq n) \\
&= \binom{N}{N-n} F^{(N-n)}(0)
\end{aligned}$$

Consider the expectation,

$$\begin{aligned}
\mathbb{E}[N_x] &= \sum_{N=1}^{\infty} NP(N_x = N) \\
&\leq \sum_{N=1}^{\infty} N \binom{N}{N-n} F^{(N-n)}(0) \\
&\leq \sum_{N=1}^{\infty} N \binom{N}{n} F^{(N-n)}(0) \\
&< \infty
\end{aligned}$$

□

(c) *Proof.* If X_i are not integer, assume $X_i \in \mathbb{R}$. If $x_k < t$, let $x_k(t) = x_k$ else, let $x_k(t) = t$. Here we let

$$N_x(t) = \sup_n \{x_1(t) + x_2(t) + \cdots + x_n(t) \leq x\}$$

$$\mathbb{E}[N_x] \leq \mathbb{E}[N_x(t)] \leq \frac{\frac{x}{t} + 1}{P(X \geq t)} < \infty$$

□

6. *Proof.* Consider $X_1 \sim \text{Cauchy}(0, n)$, $X_2 \sim \text{Cauchy}(0, 1)$

$$f_{X_1, X_2}(x_1, x_2) = \frac{1}{n\pi^2 \left(1 + \left(\frac{x_1}{n}\right)^2\right) (1 + x_2^2)}$$

$$f_{X_1+X_2, X_2}(x_1 + x_2, x_2) = \frac{1}{n\pi^2 \left(1 + \left(\frac{x_1+x_2}{n}\right)^2\right) (1 + x_2^2)}$$

$$\begin{aligned}
f_{X_1+X_2}(x_1 + x_2) &= \int f_{X_1+X_2, X_2}(x_1 + x_2, x_2) dx_2 \\
&= \int \frac{1}{n\pi^2 \left(1 + \left(\frac{x_1+x_2}{n}\right)^2\right) (1 + x_2^2)} \\
&= \int \frac{n}{\pi^2 \left(n^2 + (x_1 + x_2)^2\right) (1 + x_2^2)} \\
&= \frac{1}{(n+1)\pi \left(1 + \left(\frac{X_1+X_2}{n+1}\right)\right)}
\end{aligned}$$

This means Cauchy distribution is additive.

$$\begin{aligned}
S_n &= X_1 + X_2 + \cdots + X_n \sim \text{Cauchy}(0, n) \\
P(S_n < x) &= \frac{1}{\pi} \arctan\left(\frac{x}{n}\right) + \frac{1}{2} \\
P(S_n < nx) &= \frac{1}{\pi} \arctan(x) + \frac{1}{2} \\
&= P(X_1 < x)
\end{aligned}$$

So,

$$\frac{S_n}{n} \sim X_1$$

Which means

$$\frac{S_n}{c_n} = \frac{S_n}{n} \frac{n}{c_n} \sim X_1 \frac{n}{c_n}$$

For all x ,

$$\lim_{n \rightarrow +\infty} \frac{S_n(x)}{c_n} = \lim_{n \rightarrow +\infty} \frac{S_n(x)}{n} \frac{n}{c_n} = 0$$

So,

$$\frac{S_n}{c_n} \rightarrow 0 \Rightarrow \frac{S_n}{c_n} \xrightarrow{P} 0$$

□