

## Homework 6. Solutions

①

$$(a) \{X+Y=n\} = \bigcup_{m \in \mathbb{Z}} \{X=m, Y=n-m\}$$

The events in this union are disjoint. Therefore

$$\begin{aligned} P(X+Y=n) &= \sum_m P(X=m, Y=n-m) \\ &= \sum_m P(X=m) P(Y=n-m) \quad \text{by independence.} \end{aligned}$$

QED.

$$(b) P(X=m) = \begin{cases} e^{-\lambda} \frac{\lambda^m}{m!}, & \text{for } m \geq 0 \\ 0, & \text{for } m < 0 \end{cases} \quad \Bigg| \quad P(Y=n-m) = \begin{cases} e^{-\mu} \frac{\mu^{n-m}}{(n-m)!} & \text{for } n-m \geq 0 \\ 0 & \text{for } n-m < 0 \end{cases}$$

By part (a),

$$\begin{aligned} P(X+Y=n) &= \sum_{m=0}^n e^{-\lambda} \frac{\lambda^m}{m!} e^{-\mu} \frac{\mu^{n-m}}{(n-m)!} \\ &= e^{-(\lambda+\mu)} \sum_{m=0}^n \frac{\lambda^m \mu^{n-m}}{m! (n-m)!} \\ &= e^{-(\lambda+\mu)} \frac{1}{n!} \sum_{m=0}^n \binom{n}{m} \lambda^m \mu^{n-m} \\ &= e^{-(\lambda+\mu)} \frac{(\lambda+\mu)^n}{n!}. \end{aligned}$$

QED.

(c)  $X =$  number of successes in  $n$  <sup>independent</sup> trials (probability of success in each trial  $= p$ ).  
 $Y =$  number of successes in  $m$  independent trials, which are also independent from the trials in  $X$ .

Hence  $X+Y =$  number of successes in the total of  $(n+m)$  independent trials. Hence Bernoulli with parameters  $(n+m, p)$ .

(2)

We ~~need~~ <sup>with</sup>  $EX=0$ ,

We will find  $X, Y$  such that  $EX=0, EY=0, EXY=0$   
(therefore  $X, Y$  uncorrelated), but not independent.

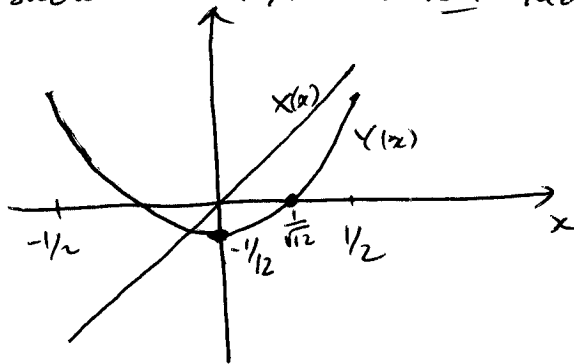
Let  $X(x) = x, Y(x) = ax^2 + b$  on  $[-1/2, 1/2]$ .

$$\text{Then } EX = \int_{-1/2}^{1/2} x dx = 0.$$

$$EY = \int_{-1/2}^{1/2} (ax^2 + b) dx = \frac{a}{12} + b$$

$$EXY = \int_{-1/2}^{1/2} x(ax^2 + b) dx = 0 \quad (\text{odd function}).$$

So if we choose  $a=1, b=-1/12$ , then  $X, Y$  are uncorrelated.  
Now we show that  $X, Y$  are not independent.



$Y(x) > 0$  for all  $x > \frac{1}{\sqrt{12}}$ .

Hence  $X > \frac{1}{\sqrt{12}}$  implies  $Y > 0$ , thus

$$P(Y > 0, X > \frac{1}{\sqrt{12}}) = P(X > \frac{1}{\sqrt{12}}). \quad (*)$$

On the other hand,  $P(Y > 0) < 1. \quad (**)$

(\*) and (\*\*) contradict the independence of  $X, Y$ .

QED

(3)

(a) let  $H(z)$  be the distribution function of  $XY$ . Then

$$H(z) = P(XY \leq z) = \int_{\mathbb{R}^2} \mathbb{1}_{\{xy \leq z\}} dP_{(x,y)}$$

where  $P_{(x,y)}$  is the joint distribution of  $X, Y$  (i.e. the distribution of the random vector  $(X, Y)$ ).

By the independence,  $P_{(x,y)}$  is the product measure of the distributions of  $P_X$  of  $X$  and  $P_Y$  of  $Y$ . So by Fubini,

$$H(z) = \int_{\mathbb{R}} \left( \int_{\mathbb{R}} \mathbb{1}_{\{xy \leq z\}} dP_X \right) dP_Y$$

||

$$\int_{\mathbb{R}} \mathbb{1}_{\{x \leq z/y\}} dP_X = P(X \leq z/y) = F(z/y).$$

Therefore,

$$H(z) = \int_{\mathbb{R}} F(z/y) dG(y).$$

(b) let  $H(z)$  be the distribution function of  $XY$ . Since  $0 \leq X \leq 1$  and  $0 \leq Y \leq 1$ ,  $H(z) = 1$  if  $z \geq 1$  and  $H(z) = 0$  if  $z \leq 0$ .

Assume then  $z \in (0, 1)$ . Then

$$H(z) = \int_0^1 F(z/y) dy \quad \text{because } H(z) = \int_{\mathbb{R}} F(z/y) g(y) dy$$

Now,  $F(x) = \begin{cases} 0, & x \leq 0 \\ x, & 0 < x \leq 1 \\ 1, & x > 1. \end{cases}$

and the density  $g(y)$  of  $Y$  is  $\mathbb{1}_{(0,1)}$ .

Hence  $H(z) = \int_0^z dy + \int_z^1 \frac{z}{y} dy = z - z \ln z.$

Answer:

$$H(z) = \begin{cases} 0, & z \leq 0 \\ z - z \ln z, & z \in (0, 1) \\ 1, & z \geq 1 \end{cases}$$

(4)

Example: consider  $n$  random variables  $X_n$  with the following distribution:

$$P(X_n = n) = \frac{1}{\sqrt{n}}; \quad P(X_n = 0) = 1 - \frac{1}{\sqrt{n}}$$

Then  $EX_n = n \cdot \frac{1}{\sqrt{n}} = \sqrt{n} \rightarrow \infty$

Therefore  $X_n \not\rightarrow 0$  in  $L^1$ .

However,  $P(X_n \neq 0) = \frac{1}{\sqrt{n}} \rightarrow 0$

Hence  $X_n \rightarrow 0$  in probability.

(5)

(a), (b): The unifying framework for both problems is a sequence of independent trials, with ~~probability~~ each successful with probability  $p$  ( $p = \frac{1}{2}$  in (a),  $p = \frac{1}{6}$  in (b))

We are interested in the expectation of the random variable

$X =$  number of trials until first success

Then 
$$P(X = k) = \underbrace{(1-p)^{k-1}}_{(k-1) \text{ failures}} \cdot \underbrace{p}_{k\text{th trial} = \text{success}}, \quad k = 1, 2, \dots$$

Hence 
$$\begin{aligned} EX &= \sum_{k=1}^{\infty} k (1-p)^{k-1} p = \sum_{k=1}^{\infty} k [(1-p)^{k-1} - (1-p)^k] \\ &= \sum_{k=1}^{\infty} k (1-p)^{k-1} - \sum_{k=1}^{\infty} k (1-p)^k = \sum_{k=0}^{\infty} (k+1) (1-p)^k - \sum_{k=0}^{\infty} k (1-p)^k \\ &= \sum_{k=0}^{\infty} (1-p)^k = \left( \frac{1}{p} \right). \end{aligned}$$

Answer in (a): 2 times

Answer in (b): 6 times

(c) Let's call the distribution of  $X$  in ~~the~~ the previous part "First success distribution with parameter  $p$ ".

Let  $X_k$ ,  $k=1, \dots, 6$ , be the number of throws until the next new face appears. So the total number of ~~times~~ throws until all faces appear is  $X_1 + \dots + X_6$ .

(Note:  $X_k$  are not independent).

Clearly,  $X_1 = 1$ .

$X_2$  has the first success distribution with parameter  $\frac{5}{6}$

$X_3$  " " " "  $\frac{4}{6}$

$X_4$  " " " "  $\frac{3}{6}$

$X_5$  " " " "  $\frac{2}{6}$

$X_6$  " " " "  $\frac{1}{6}$ .

By parts (a), (b),

$$\mathbb{E}(X_1 + \dots + X_6) = \mathbb{E}X_1 + \dots + \mathbb{E}X_6 = \frac{6}{5} + \frac{6}{4} + \frac{6}{3} + \frac{6}{2} + \frac{6}{1} = 14.7$$

(d) Similar with  $n$  instead of 6  $\Rightarrow$

$$\mathbb{E}(X_1 + \dots + X_n) = n \sum_{k=1}^n \frac{1}{k} \sim \boxed{Cn \log n}$$