Hunter Lybbert Student ID: 2426454 10-07-24 AMATH 561

## PROBLEM SET 3

1. Give an example of a probability space  $(\Omega, \mathcal{F}, P)$ , a random variable X and a function f such that  $\sigma(f(X))$  is strictly smaller than  $\sigma(X)$  but  $\sigma(f(X)) \neq \{\emptyset, \Omega\}$ . Give a function g such that  $\sigma(g(X)) = \{\emptyset, \Omega\}$ . Hint: Look at finite sample spaces with a small number of elements.

Solution:

Let our probability space be two independent coin tosses, such that  $\Omega = \{HH, TT, HT, TH\}$ . Define a random variable X such that  $X(\omega)$  be the number of heads in the outcome  $\omega$  with  $\omega \in \Omega$ . Therefore

$$X(HH) = 2,$$

$$X(TT) = 0,$$

$$X(HT) = 1, \text{ and}$$

$$X(TH) = 1.$$

Now  $\sigma(X)$  can be written as

$$\sigma(X) = \left\{ \{HH\}, \{TH, HT\}, \{TT\}, \{TT, HH\}, \{HH, HT, TH\}, \{TT, HT, TH\}, \Omega, \emptyset \right\}$$

## Part one

Random variable X and f such that  $\sigma(f(X)) \subseteq \sigma(X)$  and  $\sigma(f(X))$  is not the trivial  $\sigma$ -algebra.

Define f(x) as follows

$$f(x) = \begin{cases} 0, & x \le 0 \\ 1, & x > 0. \end{cases}$$

Let's look at the possible pre-images of f(X) with respect to a few cases of Borel sets. For convenience, I will define  $\hat{X} = f(X)$ . Now let's look at some cases for the pre-image

Case 1:  $0 \in B$  but  $1 \notin B$ 

$$\hat{X}^{-1}(B) = \left\{\omega : \hat{X}(\omega) \in B\right\} = \left\{\omega : X(\omega) \in (-\infty, 0]\right\} = \left\{TT\right\}$$

Case 2:  $0 \notin B$  but  $1 \in B$ 

$$\hat{X}^{-1}(B) = \left\{\omega: \hat{X}(\omega) \in B\right\} = \left\{\omega: X(\omega) \in (0, \infty) \right. \} = \left\{TH, HT, HH\right\}$$

Case 3:  $0 \in B$  and  $1 \in B$ 

$$\hat{X}^{-1}(B) = \left\{\omega: \hat{X}(\omega) \in B\right\} = \left\{\omega: X(\omega) \in (-\infty, \infty) \right. \} = \Omega$$

Case 4:  $0 \notin B$  and  $1 \notin B$ 

$$\hat{X}^{-1}(B) = \left\{ \omega : \hat{X}(\omega) \in B \right\} = \emptyset.$$

Therefore,

$$\sigma(f(X)) = \left\{ \{TT\}, \{TH, HT, HH\}, \Omega, \emptyset \right\} \neq \{\emptyset, \Omega\}$$

And thus we have  $\sigma(f(X)) \subseteq \sigma(X)$ .

**Part two** Now also give a function g such that  $\sigma(g(X))$  is the trivial  $\sigma$ -algebra,  $\{\emptyset, \Omega\}$ .

Define g(x) to be a constant  $c \in \mathbb{R}$  such that g(x) = c for all  $x \in \mathbb{R}$ . Once again, for convenience we define  $\tilde{X} = g(X)$ . Let's go through a few cases of what the pre-image may be for any Borel set

Case 1:  $c \in B$ 

$$\tilde{X}^{-1}(B) = \left\{ \omega : \tilde{X}(\omega) \in B \right\} = \left\{ \omega : X(\omega) \in (-\infty, \infty) \right\} = \Omega$$

Case 2:  $c \notin B$ 

$$\tilde{X}^{-1}(B) = \left\{ \omega : \tilde{X}(\omega) \in B \right\} = \emptyset.$$

Therefore,

$$\sigma(g(X)) = \{\Omega,\emptyset\} \,.$$

**2.** Give an example of events A, B, and C, each of probability strictly between 0 and 1, such that  $P(A \cap B) = P(A)P(B)$ ,  $P(A \cap C) = P(A)P(C)$ , and  $P(A \cap B \cap C) = P(A)P(B)P(C)$  but  $P(B \cap C) \neq P(B)P(C)$ . Are A, B and C independent? Hint: You can let  $\Omega$  be a set of eight equally likely points. Solution:

Let  $\Omega = \{1, 2, 3, 4, 5, 6, 7, 8\}$ . Define events A, B, and C as follows

$$A = \{1, 2, 3, 4\}$$
$$B = \{1, 2, 5, 7\}$$
$$C = \{1, 3, 6, 8\}.$$

Then we have

$$P(A \cap B) = P(\{1, 2\}) = \frac{1}{4}$$

and

$$P(A)P(B) = P(\{1, 2, 3, 4\})P(\{1, 2, 5, 7\}) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}.$$

Additionally, we have

$$P(A \cap C) = P(\{1,3\}) = \frac{1}{4}$$

and

$$P(A)P(C) = P(\{1, 2, 3, 4\})P(\{1, 3, 6, 8\}) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}.$$

Finally, we have

$$P(A \cap B \cap C) = P(\{1\}) = \frac{1}{8}$$

and

$$P(A)P(B)P(C) = P(\{1,2,3,4\})P(\{1,2,5,7\})P(\{1,3,6,8\}) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}.$$

Notice we also get

$$P(B \cap C) = P(\{1\}) = \frac{1}{8}$$

which is not equal to

$$P(B)P(C) = P(\{1, 2, 5, 7\})P(\{1, 3, 6, 8\}) = \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{4}.$$

In class we said two events E and E' are independent if  $P(E \cap E') = P(E)P(E')$ . However, since independence of a collection of events  $E_i$  for  $i \in \{1, 2, 3, ..., n\}$  implies pairwise independence,  $P(E_i \cap E_j) = P(E_i)P(E_j)$  for all  $j \neq i$ , if the collection  $E_i$  fails to be pairwise independent then the collection must not be independent either. We have shown that A and B are independent and A and C are independent. But B and C are not independent, therefore we don't have pairwise independence between each pair of the three events hence A, B, and C are not independent.  $\Box$ 

**3.** Let  $(\Omega, \mathcal{F}, P)$  be a probability space such that  $\Omega$  is countably infinite, and  $\mathcal{F} = 2^{\Omega}$ . Show that it is impossible for there to exist a countable collection of events  $A_1, A_2, \ldots \in \mathcal{F}$  which are independent, such that  $P(A_i) = 1/2$  for each i. Hint: First show that for each  $\omega \in \Omega$  and each  $n \in \mathbb{N}$ , we have  $P(\omega) \leq 1/2^n$ . Then derive a contradiction.

## Solution:

Assume by way of contradiction, there exists a countably infinite collection of independent events  $A_1, A_2, A_3, ... \in \mathcal{F}$  such that  $P(A_i) = \frac{1}{2}$ . Independence of these events implies that

$$P\left(\bigcap_{i}^{n} A_{i}\right) = \prod_{i}^{n} P(A_{i}) = \prod_{i}^{n} \frac{1}{2} = \left(\frac{1}{2}\right)^{n}.$$

I note that our collection of events is countably infinite so we can take the limit of the previous expression as  $n \to \infty$ . Their independence also implies the independence of the events  $A_i^c$ , as discussed in class. Next I want to construct a collection of new sets call them  $B_{i,j}$  where  $\omega_j \in B_{i,j}$  (note we can index the  $\omega$ 's since  $\Omega$  is countably infinite). Let  $B_{i,j}$  be

$$B_{i,j} = \begin{cases} A_i, & \omega_j \in A_i \\ A_i^c, & \omega_j \notin A_i. \end{cases}$$

Therefore we can now write each  $\omega_i$  as

$$\bigcap_{i}^{n} B_{i,j} = \{\omega_j\}.$$

Then we have

$$P(\{\omega_j\}) = P\left(\bigcap_{i}^{n} B_{i,j}\right) = \prod_{i}^{n} P(B_{i,j}) = \prod_{i}^{n} \frac{1}{2} = \left(\frac{1}{2}\right)^{n} = 0.$$

Where  $P(\{\omega_j\}) = 0$  since  $n \to \infty$  because our collection of independent events is countably infinite. Notice, since  $\Omega = \bigcup_{j=1}^{\infty} \{w_j\}$ , then

$$P(\Omega) = P(\bigcup_{j=1}^{\infty} \{w_j\}) = \sum_{j=1}^{\infty} P(\{\omega_j\}) = \sum_{j=1}^{\infty} 0 = 0.$$

Which contradicts the fact that if  $(\Omega, \mathcal{F}, P)$  is a probability space then  $P(\Omega) = 1$ . Therefore, it is impossible for there to exist a countable collection of events  $A_1, A_2, \ldots \in \mathcal{F}$  which are independent, such that  $P(A_i) = 1/2$  for each i.

**4.** (a) Let  $X \geq 0$  and  $Y \geq 0$  be independent random variables with distribution functions F and G. Find the distribution function of XY. Solution:

Let  $h(x,y) = \mathbb{1}_{\{xy \le z\}}$  and  $\mathbb{E}[h(x,y)]$  be

$$\mathbb{E}[h(x,y)] = \int_{\mathbb{R}} \int_{\mathbb{R}} h(x,y)\mu(\mathrm{d}x)\nu(\mathrm{d}y)$$

where  $\mu$  and  $\nu$  are probability measures with distribution functions F and G respectively. We also have

$$\mathbb{E}\left[h(x,y)\right] = \mathbb{E}\left[\mathbb{1}_{\{xy \le z\}}\right]$$

$$= 1 \cdot P(XY \le z) + 0 \cdot P(XY > 0)$$

$$= P(XY < z).$$

Additionally,

$$\int_{\mathbb{R}} h(x, y) \mu(\mathrm{d}x) = \int_{\mathbb{R}} \mathbb{1}_{\{xy \le z\}} \mu(\mathrm{d}x)$$
$$= \int_{\mathbb{R}} \mathbb{1}_{\{x \le \frac{z}{y}\}} \mu(\mathrm{d}x)$$
$$= P\left(X \le \frac{z}{y}\right)$$
$$= F\left(\frac{z}{y}\right).$$

Combining these we have

$$\begin{split} P\left(XY \leq z\right) &= \mathbb{E}\left[h(x,y)\right] \\ &= \int_{\mathbb{R}} \int_{\mathbb{R}} h(x,y) \mu(\mathrm{d}x) \nu(\mathrm{d}y) \\ &= \int_{\mathbb{R}} F\left(\frac{z}{y}\right) \nu(\mathrm{d}y) \\ &= \int_{\mathbb{R}} F\left(\frac{z}{y}\right) \mathrm{d}G(y) \end{split}$$

**TODO:** Account for when y is 0 somehow...

$$\int_{\mathbb{R}} F\left(\frac{z}{y}\right) dG(y) = \int_{-\infty}^{0} F\left(\frac{z}{y}\right) dG(y) + \int_{0}^{\infty} F\left(\frac{z}{y}\right) dG(y).$$

**TODO:** Consider doing this again from the beginning but in terms of x and when y is 0.

(b) If  $X \geq 0$  and  $Y \geq 0$  are independent continuous random variables with density functions f and g, find the density function of XY. Solution:

**TODO:** Redo this once you know you're starting with the right distribution function We will follow a similar process as from the class lectures

$$\int_{\mathbb{R}} F\left(\frac{z}{y}\right) \mathrm{d}G(y) = \int_{\mathbb{R}} \int_{-\infty}^{\frac{z}{y}} F\left(u\right) \mathrm{d}u \mathrm{d}G(y).$$

Now we need to do something with a change of variables along the lines of  $u = \frac{x}{y}$  then  $du = \frac{dx}{y}$ . We have

$$\begin{split} \int_{\mathbb{R}} F\left(\frac{z}{y}\right) \mathrm{d}G(y) &= \int_{\mathbb{R}} \int_{-\infty}^{\frac{z}{y}} F\left(u\right) \mathrm{d}u \mathrm{d}G(y) \\ &= \int_{\mathbb{R}} \int_{-\infty}^{z} \frac{1}{y} F\left(\frac{x}{y}\right) \mathrm{d}x \mathrm{d}G(y) \\ &= \int_{-\infty}^{z} \int_{\mathbb{R}} \frac{1}{y} F\left(\frac{x}{y}\right) \mathrm{d}G(y) \mathrm{d}x \\ &= P\left(XY \leq z\right). \end{split}$$

Therefore the density is

$$f(x) = \int_{\mathbb{R}} \frac{1}{y} F\left(\frac{x}{y}\right) dG(y).$$

(c) If X and Y are independent exponentially distributed random variables with parameter  $\lambda$ , find the density function of XY. Solution:

Recall the density function of an exponentially distributed random variable with parameter  $\lambda$  is the same as a gamma distributed random variable with parameters  $(1, \lambda)$ . Therefore the density function for one of our random variables is

$$f(x) = \frac{\lambda^1}{\Gamma(1)} x^{1-1} e^{-\lambda x} = \lambda e^{-\lambda x}.$$

**TODO:** determine if you are going to use the formula from the previous problem or if there is an easier way to think about this...