

### Homework Set No. 3 – Probability Theory (235A), Fall 2011

**Due: Tuesday 10/18/11 at discussion section**

1. Let  $X$  be an exponential r.v. with parameter  $\lambda$ , i.e.,  $F_X(x) = (1 - e^{-\lambda x})1_{[0, \infty)}(x)$ . Define random variables

$$\begin{aligned} Y &= \lfloor X \rfloor := \sup\{n \in \mathbb{Z} : n \leq x\} && \text{("the integer part of } X \text{")}, \\ Z &= \{X\} := X - \lfloor X \rfloor && \text{("the fractional part of } X \text{")}. \end{aligned}$$

(a) Compute the (1-dimensional) distributions of  $Y$  and  $Z$  (in the case of  $Y$ , since it's a discrete random variable it is most convenient to describe the distribution by giving the individual probabilities  $\mathbf{P}(Y = n)$ ,  $n = 0, 1, 2, \dots$ ; for  $Z$  one should compute either the distribution function or density function).

(b) Show that  $Y$  and  $Z$  are independent. (Hint: Check that  $\mathbf{P}(Y = n, Z \leq t) = \mathbf{P}(Y = n)\mathbf{P}(Z \leq t)$  for all  $n$  and  $t$ .)

2. (a) Let  $X, Y$  be independent r.v.'s. Define  $U = \min(X, Y)$ ,  $V = \max(X, Y)$ . Find expressions for the distribution functions  $F_U$  and  $F_V$  in terms of the distribution functions of  $X$  and  $Y$ .

(b) Assume that  $X \sim \text{Exp}(\lambda)$ ,  $Y \sim \text{Exp}(\mu)$  (and are independent as before). Prove that  $\min(X, Y)$  has distribution  $\text{Exp}(\lambda + \mu)$ . Try to give an intuitive explanation in terms of the kind of real-life phenomena that the exponential distribution is intended to model (e.g., measuring the time for a light-bulb to burn out, or for a radioactive particle to be emitted from a chunk of radioactive material).

(c) Let  $X_1, X_2, \dots$  be a sequence of independent r.v.'s, all of them having distribution  $\text{Exp}(1)$ . For each  $n \geq 1$  denote

$$M_n = \max(X_1, X_2, \dots, X_n) - \log n.$$

Compute for each  $n$  the distribution function of  $M_n$ , and find the limit (if it exists)

$$F(x) = \lim_{n \rightarrow \infty} F_{M_n}(x).$$

3. If  $X, Y$  are r.v.'s with a joint density  $f_{X,Y}$ , the identity

$$\mathbf{P}((X, Y) \in A) = \iint_A f_{X,Y}(x, y) dx dy$$

holds for all “reasonable” sets  $A \subset \mathbb{R}^2$  (in fact, for all Borel-measurable sets, but that requires knowing what that integral means for a set such as  $\mathbb{R}^2 \setminus Q^2 \dots$ ). In particular, if  $X, Y$  are independent and have respective densities  $f_X$  and  $f_Y$ , so  $f_{X,Y}(x, y) = f_X(x)f_Y(y)$ , then

$$F_{X+Y}(t) = \mathbf{P}(X + Y \leq t) = \int_{-\infty}^{\infty} \int_{-\infty}^{t-x} f_X(x)f_Y(y) dy dx.$$

Differentiating with respect to  $t$  gives (assuming without justification that it is allowed to differentiate under the integral):

$$f_{X+Y}(t) = \int_{-\infty}^{\infty} f_X(x)f_Y(t-x) dx.$$

Use this formula to compute the distribution of  $X + Y$  when  $X$  and  $Y$  are independent r.v.'s with the following (pairs of) distributions:

1.  $X \sim U[0, 1], Y \sim U[0, 2]$ .
2.  $X \sim \text{Exp}(1), Y \sim \text{Exp}(1)$ .
3.  $X \sim \text{Exp}(1), -Y \sim \text{Exp}(1)$ .

4. (a) Let  $(A_n)_{n=1}^{\infty}$  be a sequence of events in a probability space. Show that

$$1_{\limsup A_n} = \limsup_n 1_{A_n}.$$

(The lim-sup on the left refers to the lim-sup operation on events; on the right it refers to the lim-sup of a sequence of functions; the identity is an identity of real-valued functions on  $\Omega$ , i.e., should be satisfied for each individual point  $\omega \in \Omega$  in the sample space). Similarly, show (either separately or by relying on the first claim) that

$$1_{\liminf A_n} = \liminf_n 1_{A_n}.$$

(b) Let  $U$  be a uniform random variable in  $(0, 1)$ . For each  $n \geq 1$  define an event  $A_n$  by

$$A_n = \{U < 1/n\}.$$

Note that  $\sum_{n=1}^{\infty} \mathbf{P}(A_n) = \infty$ . However, compute  $\mathbf{P}(A_n \text{ i.o.})$  and show that the conclusion of the second Borel-Cantelli lemma does not hold (of course, one of the assumptions of the lemma also doesn't hold, so there's no contradiction).

**5.** If  $P, Q$  are two probability measures on a measurable space  $(\Omega, \mathcal{F})$ , we say that  $P$  is **absolutely continuous with respect to**  $Q$ , and denote this  $P \ll Q$ , if for any  $A \in \mathcal{F}$ , if  $Q(A) = 0$  then  $P(A) = 0$ .

Prove that  $P \ll Q$  if and only if for any  $\epsilon > 0$  there exists a  $\delta > 0$  such that if  $A \in \mathcal{F}$  and  $Q(A) < \delta$  then  $P(A) < \epsilon$ .

**Hint.** Apply a certain famous lemma.

**Note.** The intuitive meaning of the relation  $P \ll Q$  is as follows: suppose there is a probabilistic experiment, and we are told that one of the measures  $P$  or  $Q$  governs the statistical behavior of the outcome, but we don't know which one. (This is a situation that arises frequently in real-life applications of probability and statistics.) All we can do is perform the experiment, observe the result, and make a guess. If  $P \ll Q$ , any event which is observable with positive probability according to  $P$  also has positive  $Q$ -probability, so we can never rule out  $Q$  as the correct measure, although we may get an event with  $Q(A) > 0$  and  $P(A) = 0$  that enables us to rule out  $P$ . If we also have the symmetric relation  $Q \ll P$ , then we can't rule out either of the measures.