## From the textbook:

Chapter 6 / Problems: 6.10, 6.19, 6.38, 6.40, 6.41, 6.43, 6.48; Theoretical exercises: 6.14

## Additional problems:

- **A.10.1.** If  $X_1$  and  $X_2$  are independent geometric(p) random variables, show that  $X_1 + X_2$  has a negative-binomial(2, p) distribution: recall for integer  $r \ge 1$  a discrete rv W has the negative-binomial(r, p) provided the pmf of W is  $p(w) = {w-1 \choose r-1} p^r (1-p)^{w-r}$  for  $w = r, r+1, r+2, \ldots$
- **A.10.2.** An amplifier uses a component whose time until failure has an exponential distribution with mean  $\beta$ , i.e, pdf  $f(x) = \frac{1}{\beta}e^{-x/\beta}$  for x > 0. Once this component fails it is immediately (and instantaneously) replaced by an identical component with identical lifetime distribution. If  $X_i$  represents the lifetime of the *i*th component, and we have a total of n replacements (including the one already in the amplifier), what is the distribution of the total lifetime  $\sum_{i=1}^{n} X_i$  of the amplifier? Hint: first find the distribution of  $Y_1 = X_1 + X_2$ . Then find the distribution of  $Y_1 + X_3$ , etc. Find a pattern.
- **A.10.3.** (continued) A special case of **A.10.2** has if X, Y are independent  $\exp(1)$ , then X + Y has a  $\operatorname{Gamma}(2,1)$ . If u>0 is given, find the conditional pdf  $f_{X|X+Y}(x|u)$  of X given X+Y=u. Hint:  $f_{X|X+Y}(x|u)=\frac{f_{X,Y}(x,u-x)}{f_{X+Y}(u)}\dots$  do you see why?
- **A.10.4.** (The multinomial distribution) We generalize the binomial distribution: Suppose we have n independent and identical trials where each trial can result in any one of r possible outcomes, the probability a trial results in outcome i (i = 1, 2, ..., r) is  $p_i$  which is the same from trial to trial (think of rolling an r-sided die n times). If we let  $X_i$  count the number of trials that result in outcome i, then the vector  $(X_1, X_2, ..., X_r)$  has the so-called multinomial distribution:

$$P(X_1 = x_1, X_2 = x_2, \dots, X_r = x_r) = \frac{n!}{x_1! x_2! \cdots x_r!} p_1^{x_1} p_2^{x_2} \cdots p_r^{x_r},$$

 $x_i \ge 0$  are integer,  $\sum_{i=1}^r x_i = n$  and  $\sum_{i=1}^r p_i = 1$ . Three important facts about the multinomial are that (1) the above is a pmf - this is a multinomial theorem, (2) the random variables  $X_i$  are dependent, and (3) for each i,  $X_i$  has a binomial  $(n, p_i)$  distribution. A hard way of showing fact (3) is to compute the marginal of  $X_i$  by summing out all possible values of  $x_j$  for  $j \ne i$ . Find a much easier explanation for why  $X_i$  is binomial  $(n, p_i)$ .

- **A.10.5.** (a) Suppose  $X_1, X_2, \ldots, X_n$  are independent uniform (0, 1) random variables. Let  $Y_n := \max\{X_1, X_2, \ldots, X_n\}$ . Show that the cdf of  $Y_n$  is  $F_n(y) = y^n$  for  $0 \le y \le 1$ .
- (b) Suppose  $X_1, X_2, \ldots, X_n$  are independent uniform (0,1) rvs. Let  $Y_1 := \min\{X_1, X_2, \ldots, X_n\}$ . Show that the cdf of  $Y_1$  is  $F_1(y) = 1 (1 y)^n$  for  $0 \le y \le 1$ .
- (c) Compute  $E(Y_n)$  and  $E(Y_1)$ .
- (d)\* Suppose  $X_1, X_2, X_3, \ldots, X_n, \ldots$  is an i.i.d. sequence of uniform (0,1) rvs. For each fixed n, find a formula for the cdf of  $W_n := n \min\{X_1, \ldots, X_n\}$ , that is,  $G_n(w) = P(n \min\{X_1, \ldots, X_n\} \le w)$  for 0 < w < n. Then show that for any w > 0, that  $\lim_{n \to \infty} G_n(w) = 1 e^{-w}$  and conclude that as n gets large the random variables  $W_n$  have a distribution that is converging to a unit exponential.
- **A.10.6.** Consider the jointly continuous rvs X, Y having joint pdf  $f(x,y) = e^{-y}$  for  $0 < x < y < \infty$ .
- (a) Compute the marginal pdf of X and the marginal pdf of Y.
- (b) Determine whether X, Y are independent or not.
- (c) Compute the conditional pdf of X given Y = y, and the conditional pdf of Y given X = x.
- (d) Evaluate P(Y > 2|X = 1). Also evaluate P(Y > 2|X > 1).