

WEEK 10 EXERCISES

You are expected to do all these problems, but for **Homework 10** please turn in **only Problems 3, 5, and 7** on **Thursday November 1 at the start of lecture**.

1. A Mixed Population

A population consists of men, women, and children. Here are some summary statistics on their heights.

- 100 women, average height 65 inches, SD of heights 3 inches
- 75 men, average height 69 inches, SD of heights 4 inches
- 25 children, average height 50 inches, SD of heights 2 inches

Let X be the height of a person picked at random from this population. Find $E(X)$ and $SD(X)$.

2. Heads in Tosses of a Random Coin

A random variable X with values in $(0, 1)$ has expectation μ and SD σ .

Fix a positive integer n . Given that $X = p$, toss a p -coin n times. Let H be the number of heads.

Find $E(H)$ and $Var(H)$.

3. Two-Colored Die

A die that has 2 blue faces and 4 green faces is rolled repeatedly. Let R be the number of rolls till both colors have appeared. Find $E(R)$ and $Var(R)$.

4. Comparing Waiting Times

A p -coin is tossed repeatedly. Let W_H be the number of tosses till the first H, W_{HH} the number of tosses till the first time the pattern HH appears, and W_{HT} the number of tosses till the first time the pattern HT appears.

(a) Let $p = 1/2$. Explain *without calculation* why $E(W_{HH})$ must be different from $E(W_{HT})$, and say which of the two values is smaller. [Think about what happens when the pattern is *not* completed on the toss following W_H .]

(b) Now let $0 < p < 1$. By conditioning appropriately, find $E(W_{HT})$ in terms of p .

(c) Find $E(W_{HHH})$ and simplify it as much as possible. Then guess an answer for the expected number of tosses till you get n heads in a row.

5. Overlapping Tosses

Consider a sequence of i.i.d. Bernoulli (p) trials. Consider the three variables X , Y , and Z defined by:

- X is the number of successes in trials 1 through 100
- Y is the number of successes in trials 51 through 100

- Z is the number of successes in trials 51 through 150

- For each of X , Y , and Z , say what the distribution is and provide the parameters.
- Fix k in the range $0, 1, \dots, 100$ and find the conditional distribution of Y given $X = k$. Recognize this as a famous one and provide the parameters.
- Find the least squares predictor of Y based on X and say whether it is a linear function of X . (If it is, then the best linear predictor is in fact the best among all predictors.)
- Find $\text{Var}(Y | X)$.
- Find $E(Z | X)$ and $\text{Var}(Z | X)$.

6. Predicting Max Based on Min

Let U_1, U_2, \dots, U_n be i.i.d. uniform $(0, 1)$ variables. As usual, let $U_{(1)}$ be their minimum and $U_{(n)}$ their maximum.

- Draw n points on the unit interval, and use the diagram to guess the least squares predictor of $U_{(n)}$ based on $U_{(1)}$. Symmetry is your friend.
- Prove that your guess is correct.
- Is the best predictor in Parts (a) and (b) a linear function of $U_{(1)}$? (If it is, then the best linear predictor is in fact the best among all predictors.)

7. Mixtures and Moment Generating Functions

Start with two random variables X and Y , and construct the new random variable W as follows:

Toss a p -coin. If it lands heads, let $W = X$. If it lands tails, let $W = Y$.

- Write the cdf F_W of W in terms of the cdf F_X of X and the cdf F_Y of Y . The answer will show why the distribution of W is called a *mixture* of the distributions of X and Y . It is important to keep in mind that the value of W is either a value of X or a value of Y ; it's not an average of the two. But the distribution of W is an average of the two distributions.
- Write the mgf M_W of W in terms of the mgf M_X of X and the mgf M_Y of Y . You can assume that all the mgf's exist in an interval around 0, which is the condition needed for all the nice properties of mgf's to hold.
- Suppose X is standard normal and Y is normal with mean $\mu > 0$ and variance 1. Could the distribution of W be normal? Why or why not?
- Let V have cdf F_V defined by $F_V(v) = 1 - \frac{e^{-v}}{5}$ for $v \geq 0$ and 0 otherwise. Sketch (carefully) the graph of F_V . Then describe the distribution of V as a mixture of two distributions whose moment generating functions you know, and hence find M_V .

8. MGF of a Random Sum

Let N be a non-negative integer valued random variable, and let X, X_1, X_2, \dots be i.i.d. and independent of N . As in class, define the "random sum" S by

$$\begin{aligned} S &= 0 \text{ if } N = 0 \\ &= X_1 + X_2 + \dots + X_n \text{ if } N = n > 0 \end{aligned}$$

(a) Let M be our usual notation for moment generating functions. By conditioning on N , show that

$$M_S(t) = M_N(\log M_X(t))$$

assuming that all the quantities above are well defined. [The identity $w = e^{\log(w)}$ might be handy.]

(b) Let N have the geometric (p) distribution on $\{1, 2, 3, \dots\}$. Find the mgf of N . This doesn't use Part (a).

(c) Let X_1, X_2, \dots be i.i.d. exponential (λ) variables and let N be geometric as in Part (b). Use the results of Parts (a) and (b) to identify the distribution of S .

(d) Find the density of S by conditioning on N , and hence confirm the result of Part (c).

[Find $P(S \in ds)$ by conditioning on N .]

(e) Use a Poisson (λ) process and Poissonization (also known as “thinning” in the context of Poisson processes) to find yet another way of confirming the result of Part (c).