Stat 345/445: Theoretical Statistics I: Homework 8 Solutions

Textbook Exercises

5.3 (345: 1 pt.) Let X_1, \ldots, X_n be iid random variables with continuous cdf F_X , and suppose $\mathbb{E}X_i = \mu$. Define the random variables Y_1, \ldots, Y_n by

$$Y_i = \begin{cases} 1 & \text{if } X_i > \mu \\ 0 & \text{if } X_i \le \mu. \end{cases}$$

Find the distribution of $\sum_{i=1}^{n} Y_i$.

Note that $Y_i \sim \text{Bernoulli}$ with $p_i = P(X_i \geq \mu) = 1 - F(\mu)$ for each i. Since the Y_i 's are iid Bernoulli,

$$\sum_{i=1}^{n} Y_i \sim \text{binomial}(n, p = 1 - F(\mu)).$$

5.4 (445: 2 pts.) A generalization of iif random variables is *exchangeable* random variables, an idea due to deFinetti (1972). A discussion of exchangeability can also be found in Feller (1971). The random variables X_1, \ldots, X_n are *exchangeable* if any permutation of any subset of them of size k ($k \le n$) has the same distribution. In this exercise we will see an example of random variables that are exchangeable but not iid. Let $X_i | P \sum$ iid Bernouli(P), $i = 1, \ldots, n$, and let $P \sim$ uniform (0,1).

(a) Show that the marginal distribution of any k of the Xs is the same as

$$P(X_1 = x_1, \dots, X_k = x_k) = \int_0^1 p^t (1-p)^{k-t} dp = \frac{t!(k-t)!}{(k+1)!},$$

where $t = \sum_{i=1}^{k} x_i$. Hence, the Xs are exchangeable.

$$X_i|P \sim \text{Bernoulli}(P).$$
 $P \sim \text{uniform}(0,1).$

$$P(X_1 = x_1, \dots, X_k = x_k | P) = P^{\sum_{i=1}^k x_i} (1-p)^{k-\sum_{i=1}^k x_i} = p^t (1-p)^{k-t}$$

where $t = \sum_{i=1}^{k} x_i$.

$$P(X_1 = x_1, \dots, X_k = x_k) = \int_{-\infty}^{\infty} P(X_1 = x_1, \dots, X_k = x_k | P) f(p) dp$$

$$= \int_0^1 p^t (1 - p)^{k - t} dp$$

$$= \beta(t + 1, k - t + 1)$$

$$= \frac{\Gamma(t + 1)\Gamma(k - t + 1)}{\Gamma(k + 2)}$$

$$= \frac{t!(k - t)!}{(k + 1)!}$$

(b) Show that, marginally,

$$P(X_1 = x_1, \dots, X_n = x_n) \neq \prod_{i=1}^n P(X_i = x_i),$$

so the distribution of the Xs is exchangeable but not iid.

$$P(X_i = x_i) = \frac{x_i!(1 - x_i)!}{2!} = \begin{cases} \frac{1}{2} & x_i = 0\\ \frac{1}{2} & x_i = 1 \end{cases}$$

$$P(X_1 = x_1, X_2 = x_2, \dots, X_n = x_n) = (\frac{1}{2})^n \neq \frac{t!(n - t)!}{(n + 1)!} \implies \text{not iid}$$

- **5.6** If X has pdf $f_X(x)$ and Y, independent of X, has pdf $f_Y(y)$, establish formulas, similar to (5.2.3), for the random variables Z in each of the following situations.
 - (a) (345 & 445: 1 pt.) Z = X Y

For
$$Z = X - Y$$
, set $W = X$. Then $Y = W - Z$, $X = W$, and $|J| = \begin{vmatrix} 0 & 1 \\ -1 & 1 \end{vmatrix} = 1$. Then $f_{Z,W}(z,w) = f_X(w)f_Y(w-z) \cdot 1$, thus $f_Z(z) = \int_{-\infty}^{\infty} f_X(w)f_Y(w-z)dw$.

(b) (345 & 445: 1 pt.) Z = XY

For
$$Z = XY$$
, set $W = X$. Then $Y = Z/W$ and $|J| = \begin{vmatrix} 0 & 1 \\ 1/w & -z/w^2 \end{vmatrix} = -1/w$. Then $f_{Z,W}(z,w) = f_X(w)f_Y(z/w) \cdot |-1/w|$, thus $f_Z(z) = \int_{-\infty}^{\infty} |-1/w|f_X(w)f_Y(z/w)dw$.

- **5.10** Let X_1, \ldots, X_n be a random sample from a $n(\mu, \sigma^2)$ population.
 - (a) (345 & 445: 1.5 pts.) Find expressions for $\theta_1, \ldots, \theta_4$, as defined in Exercise 5.8, in terms of μ and σ^2 .

$$\theta_{1} = \mathbb{E}X_{i} = \mu$$

$$\theta_{2} = \mathbb{E}(X_{i} - \mu)^{2} = \sigma^{2}$$

$$\theta_{3} = \mathbb{E}(X_{i} - \mu)^{3} = \mathbb{E}(X_{i} - \mu)^{2}(X_{i} - \mu) \qquad (Stein's lemma: \mathbb{E}g(X)(X - \theta) = \sigma^{2}\mathbb{E}g'(X))$$

$$= 2\sigma^{2}\mathbb{E}(X_{i} - \mu) = 0$$

$$\theta_{4} = \mathbb{E}(X_{i} - \mu)^{4} = \mathbb{E}(X_{i} - \mu)^{4} = \sigma^{2}\mathbb{E}(X_{i} - \mu) = \sigma^{2}\mathbb{E}(X_{i} - \mu) = \sigma^{2}\mathbb{E}(X_{i} - \mu)^{2} = \sigma^{2}\mathbb{E}(X_{i} - \mu)^{2}$$

(b) (345: 1 pt. & 445: 0.5 pt.) Use the results from Exercise 5.8, together with the results of part (a), to calculate $\text{Var } S^2$.

$$VarS^{2} = \frac{1}{n} (\theta_{4} - \frac{n-3}{n-1} \theta_{2}^{2}) = \frac{1}{n} (3\sigma^{4} - \frac{n-3}{n-1} \sigma^{4}) = \frac{2\sigma^{4}}{n-1}.$$

(c) (345: 1 pt. & 445: 0.5 pt.) Calculate Var S^2 a completely different (and easier) way: Use the fact that $(n-1)S^2/\sigma^2 \sim \chi^2_{n-1}$.

Use the fact that $(n-1)S^2/\sigma^2 \sim \chi^2_{n-1}$ and $\mathrm{Var}\chi^2_{n-1} = 2(n-1)$ to get

$$\operatorname{Var}\left(\frac{(n-1)S^2}{\sigma^2}\right) = 2(n-1)$$

which implies $\left(\frac{(n-1)^2}{\sigma^4}\right) \text{Var} S^2 = 2(n-1)$ and hence

$$VarS^{2} = \frac{2(n-1)}{(n-1)^{2}/\sigma^{4}} = \frac{2\sigma^{4}}{n-1}.$$

5.15 (345 & 445: 2 pts.) Establish the following recursion relations for means and variances. Let \bar{X}_n and S_n^2 be the mean and variance, respectively, of X_1, \ldots, X_n . Then suppose another observation, X_{n+1} , becomes available. Show that

(a)
$$\bar{X}_{n+1} = \frac{X_{n+1} + n\bar{X}_n}{n+1}$$

$$\bar{X}_{n+1} = \frac{\sum_{i=1}^{n+1} X_i}{n+1} = \frac{X_{n+1} + \sum_{i=1}^{n} X_i}{n+1} = \frac{X_{n+1} + n\bar{X}_n}{n+1}.$$

(b)
$$nS_{n+1}^2 = (n-1)S_n^2 + (\frac{n}{n+1})(X_{n+1} - \bar{X}_n)^2$$
.

$$nS_{n+1}^{2} = \frac{n}{(n+1)-1} \sum_{i=1}^{n+1} (X_{i} - \bar{X}_{n+1})^{2}$$

$$= \sum_{i=1}^{n+1} (X_{i} - \frac{X_{n+1} + n\bar{X}_{n}}{n+1})^{2} \qquad \text{use (a)}$$

$$= \sum_{i=1}^{n+1} (X_{i} - \frac{X_{n+1}}{n+1} - \frac{n\bar{X}_{n}}{n+1})^{2}$$

$$= \sum_{i=1}^{n+1} \left[(X_{i} - \bar{X}_{n}) - \left(\frac{X_{n+1}}{n+1} - \frac{\bar{X}_{n}}{n+1} \right) \right]^{2}$$

$$= \sum_{i=1}^{n+1} \left[(X_{i} - \bar{X}_{n})^{2} - 2(X_{i} - \bar{X}_{n}) \left(\frac{X_{n+1} - \bar{X}_{n}}{n+1} \right) + \frac{1}{(n+1)^{2}} \left(X_{n+1} - \bar{X}_{n} \right)^{2} \right]$$

$$= \sum_{i=1}^{n} (X_{i} - \bar{X}_{n})^{2} + (X_{n+1} - \bar{X}_{n})^{2} - 2\frac{(X_{n+1} - \bar{X}_{n})^{2}}{n+1} + \frac{n+1}{(n+1)^{2}} (X_{n+1} - \bar{X}_{n})^{2}$$

$$= (n-1)S_{n}^{2} + \frac{n}{n+1} (X_{n+1} - \bar{X}_{n})^{2}.$$

Since $\sum_{1}^{n} (X_i - \bar{X}_n) = 0$.

5.16 (345 & 445: 1.5 pts.) Let X_i , i = 1, 2, 3, be independent with $n(i, i^2)$ distributions. For each of the following situations, use the X_i s to construct a statistic with the indicated distribution.

(a) chi squared with 3 degrees of freedom

$$\sum_{i=1}^{3} (\frac{X_i - i}{i})^2 \sim \chi_3^2$$

(b) t distributions with 2 degrees of freedom

$$(\frac{X_i - 1}{i}) / \sqrt{\sum_{i=2}^{3} (\frac{X_i - i}{i})^2 / 2} \sim t_2$$

(c) F distribution with 1 and 2 degrees of freedom

Square the random variable in part (b).