

## STAT 512 su 2021 hw 3

1. Let  $X_1, \dots, X_{25} \stackrel{\text{ind}}{\sim} \text{Normal}(\mu = 3, \sigma^2 = 4)$ .
  - (a) Give the mgf of  $X_1$ .
  - (b) Give the mgf of  $\bar{X}_{25} = (1/25)(X_1 + \dots + X_{25})$ .
  - (c) Give  $P(X_1 < 2)$ .
  - (d) Give  $P(\bar{X}_{25} < 2)$ .
  - (e) Give  $P(|X_1 - 3| > 1)$ .
  - (f) Give  $P(|\bar{X}_{25} - 3| > 1)$ .
  - (g) Identify the distribution of  $5(\bar{X}_{25} - 3)/2$ .
  - (h) Give  $P([5(\bar{X}_{25} - 3)/2]^2 > 3.841459)$ . *Hint: Use the result from Question 2.*
  - (i) Now let  $X_1, \dots, X_n \stackrel{\text{ind}}{\sim} \text{Normal}(\mu = 3, \sigma^2 = 4)$  for  $n \geq 1$  and set  $\bar{X}_n = (1/n)(X_1 + \dots + X_n)$ .
    - i. Consider the probability  $P(|\bar{X}_n - 3| > \varepsilon)$ , for some small  $\varepsilon > 0$ . Is this an increasing or a decreasing function of  $n$ ?
    - ii. What does your answer say about the quality of  $\bar{X}_n$  as an estimator of the mean?
2. Let  $Z \sim \text{Normal}(0, 1)$ . Show that  $Z^2$  has the  $\chi_1^2$  distribution (chi-squared with degrees of freedom 1), i.e. show that the pdf of  $Y = Z^2$  is

$$f_Y(y) = \frac{1}{\Gamma(1/2)2^{1/2}} y^{\frac{1}{2}-1} e^{-y/2} \mathbf{1}(y > 0).$$

*Hint:  $\Gamma(1/2) = \sqrt{\pi}$ .*

3. Let  $W \sim \text{Chi-squared}(\nu)$ , so that the pdf of  $W$  is given by

$$f_W(w) = \frac{1}{\Gamma(\nu/2)2^{\nu/2}} w^{\nu/2-1} e^{-w/2} \mathbf{1}(w > 0),$$

where  $\nu > 0$  is the degrees of freedom.

- (a) Give  $\alpha$  and  $\beta$  such the  $\text{Gamma}(\alpha, \beta)$  and  $\text{Chi-squared}(\nu)$  distributions are the same.
  - (b) Give  $\mathbb{E}W$  in terms of the degrees of freedom parameter  $\nu$ .
  - (c) Give  $\text{Var } W$  in terms of the degrees of freedom parameter  $\nu$ .
4. Let  $Y_1, \dots, Y_n \stackrel{\text{ind}}{\sim} \text{Bernoulli}(p)$  distribution,  $p \in (0, 1)$ . Let  $\hat{p}_n = \bar{Y}_n = n^{-1}(Y_1 + \dots + Y_n)$ .
  - (a) Show that

$$\frac{1}{n-1} \sum_{i=1}^n (Y_i - \bar{Y}_n)^2 = \frac{n}{n-1} \hat{p}_n(1 - \hat{p}_n).$$

- (b) Find  $\mathbb{E}[n(n-1)^{-1} \hat{p}_n(1 - \hat{p}_n)]$ .

5. Let  $X_1, \dots, X_n \stackrel{\text{ind}}{\sim} \text{Poisson}(\lambda)$  distribution. Find the expected value of  $\hat{\lambda} = n^{-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2$ .  
*Hint:  $\mathbb{E}\hat{\lambda} \neq \lambda$ .*

6. Let  $X_1, \dots, X_n$  be a random sample from the  $\text{Weibull}(k, \lambda)$  distribution, which has pdf

$$f(x) = \begin{cases} \frac{k}{\lambda} \left(\frac{x}{\lambda}\right)^{k-1} e^{-(x/\lambda)^k}, & x \geq 0 \\ 0, & x < 0, \end{cases}$$

for  $k, \lambda > 0$ .

(a) Find the cdf  $F_X$  of the  $\text{Weibull}(k, \lambda)$  distribution. *Hint: Set up the integral  $F_X(x) = \int_0^x f_X(t)dt$  and do the change of variable  $u = (t/\lambda)^k$ .*

(b) Find the pdf of  $X_{(1)}$ .

(c) Show that  $X_{(1)}$  has the  $\text{Weibull}(k, \lambda n^{-1/k})$  distribution.

7. Let  $U_1, \dots, U_n$  be a random sample from the  $\text{Uniform}(0, \theta)$  distribution.

(a) Find the joint density of order statistics  $U_{(1)}$  and  $U_{(n)}$ .

(b) Find the joint density of the random variables  $R = U_{(1)}/U_{(n)}$  and  $M = U_{(n)}$ .

(c) State whether  $R$  and  $M$  are independent.

(d) Give the marginal pdf of  $R$  and identify the distribution.

8. Let  $X_1, \dots, X_n$  be a random sample from the  $\text{Uniform}(0, 1)$  distribution, where  $n$  is an odd number. Show that

(a) The expected value of the median is  $1/2$ .

(b) The variance of the median is  $\frac{1}{4(n+2)}$ .

9. Use R to run the following simulation. Choose a sample size  $n \leq 20$  and draw 1,000 samples of size  $n$  from the  $\text{Uniform}(0, 1)$  distribution. In so doing:

(a) Choose a value of  $k$ ,  $1 < k < n$ , and from each of the 1,000 samples, save the  $k$ th order statistic. Make a histogram of the 1,000 values of the  $k$ th order statistic and overlay the pdf of the sampling distribution of the  $k$ th order statistic (you must figure out and input the shape parameters of the beta distribution). Use the following code to get started:

```
S <- 1000 # number of random samples to generate
Yk <- numeric(S) # create empty vector in which to store values

for(s in 1:S) # run a loop of length S
{
  Y <- runif(n)
  Yk[s] <- sort(Y)[k] # get kth order statistic
}

hist(Yk, freq=FALSE, xlim=c(0,1))
y.seq <- seq(0,1, length=100)
lines(dbeta(y.seq, shape1=???, shape2=???)~y.seq, col="blue", lwd=2)
```

(b) Do the same thing for the  $n$ th order statistic and the 1st order statistic.

Turn in all three histograms with densities overlaid and all your code.