## Homework 5

By signing my name, I agree to abide by the Stern Code of Conduct \_\_\_\_\_

## Question 1

This question is about the sampling distribution of the mean, and using  $\bar{X}$  as an estimator for the parameter  $\mu$ . Suppose  $X_1, \ldots, X_n$  are independent and identically distributed (i.i.d.) with the same distribution as X.

a. Recall that  $E[\bar{X}] = E[X]$  and  $\text{Var}(\bar{X}) = \frac{\text{Var}(X)}{n}$ . Do we need to assume anything more about X for  $\bar{X}$  to be an unbiased estimate of  $\mu$ , and if so, what?

**Solution**: Since the bias is  $E[\bar{X} - \mu] = E[\bar{X}] - \mu = E[X] - \mu$ , we need to assume  $E[X] = \mu$ .

b. One way to write Chebyshev's inequality is  $P(|Y - E[Y]| \ge k\sqrt{\text{Var}(Y)}) \le 1/k^2$ . Let  $Y = \bar{X}$  and write this inequality out in terms of  $\bar{X}, E[X], k, \text{Var}(X)$ , and n.

Solution:

$$P\left(|\bar{X} - E[X]| \ge k\sqrt{\frac{\operatorname{Var}(X)}{n}}\right) \le \frac{1}{k^2}$$

c. If we let  $k = \sqrt{n}$ , the answer from the previous part should simplify to  $P(|\bar{X} - E[X]| \ge \sqrt{\text{Var}(X)}) \le 1/n$ . If we want 99% probability that the sample mean is *within one standard deviation* of X away from E[X], how large of a sample do we need?

**Solution**: If 1/n = 1/100 then there is a 1 - 1/100 = 99% probability of  $\bar{X}$  being within  $\sigma$  from E[X]. So we need n = 100.

d. Suppose  $\bar{X}$  is unbiased:  $E[\bar{X}] = \mu$ . Compute the mean squared error of this estimate,  $MSE(\bar{X})$ .

**Solution**: Using the bias-variance decomposition, since we know the Bias $(\bar{X})^2$  term is 0 (unbiased), we have  $MSE(\bar{X}) = Var(\bar{X}) = \sigma^2/n$ .

e. Assume now that  $X \sim \text{Ber}(p)$ . For some constant c,  $c\bar{X}$  has a distribution that you know. What is the constant, and what is the distribution?

**Solution**: If c = n then  $n\bar{X} = \sum_{i=1}^{n} X_i$  is a sum of n independent Bernoulli random variables with success probability p, hence  $n\bar{X} \sim \text{Bin}(n,p)$ .

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## Question 2

This question will guide you through a *simulation study* in R to understand the bias of a certain estimator. Suppose  $U_1, \ldots, U_n$  are independent and identically distributed as  $U[0, \theta]$ , with  $\theta = 1$  and let  $\hat{\theta} = \max\{U_1, \ldots, U_n\}$ . Since we are generating this data ourselves, we know the true value of  $\theta = 1$ , so we can compute the bias of  $\hat{\theta}$ . Our goal will be to study how this bias decreases as the sample size n increases.

The following code creates a function that you can use to generate observations of  $\hat{\theta}$ .

```
theta_hat <- function(n) return(max(runif(n)))</pre>
```

You need to run this code once so that R learns the definition of theta\_hat. After that, you can "call" the function by running, for example, theta\_hat(10) to generate one observation of a maximum of sample size n = 10.

```
theta_hat(10)
```

```
## [1] 0.9207211
```

To generate a sample of many i.i.d. copies of  $\hat{\theta}$ , we use the replicate function:

```
replicate(10, theta_hat(10))
```

```
## [1] 0.9004413 0.9898925 0.9436040 0.9708578 0.8303971 0.8675335 0.7628322 ## [8] 0.9685146 0.8851466 0.9208831
```

Finally, we estimate the bias by generating many samples of  $\hat{\theta}$ , taking their average, and subtracting  $\theta$ :

```
mean(replicate(10000, theta_hat(10))) - 1
```

```
## [1] -0.09043724
```

a. Run this code again to estimate the bias when  $\hat{\theta}$  is based on a sample of size n=100, and again for a sample of size n=1000.

Solution:

```
c(mean(replicate(10000, theta_hat(100))) - 1,
mean(replicate(10000, theta_hat(1000))) - 1)
```

```
## [1] -0.009974618 -0.001023195
```

b. Compute these answers to the answer I gave in class: compute (n-1)/n - 1 = -1/n for the same values of n. Are the simulation estimates of bias reasonably close to the exact mathematical answer?

**Solution**: -1/100 = -0.01, -1/1000 = -0.001. Yes, the values from the simulation were fairly close to these.

c. Use the sd function to estimate the standard deviation of  $\hat{\theta}$  based on n=10 and on n=100.

```
c(sd(replicate(10000, theta_hat(10))),
sd(replicate(10000, theta_hat(100))))
```

```
## [1] 0.083129768 0.009785584
```