## Homework 1 Exercise 3

Chutian Chen cc4515; Congcheng Yan cy2550; Mingrui Liu ml4404

February 11, 2020

## Exercise 3

1. Let Y has a standard normal distribution. Let X has a normal distribution  $N\left(\mu,\sigma^2\right)$ . We have

$$\sigma^2 = E(X^2) - (E(X))^2,$$
  
 $E(X^2) = \sigma^2 + \mu^2.$ 

Then we have some conclusions about Y:

$$E(Y) = 0,$$
 
$$E(Y^2) = 0 + 1 = 0,$$
 
$$E(Y^3) = \int_{-\infty}^{\infty} x^3 \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx = 0.$$

Go back to  $\gamma$ ,

$$\begin{split} \gamma &= \mathbb{E}\left[R_1^3\right] \\ &= E[(Y*\sigma + \mu)^3] \\ &= E[Y^3*\sigma^3 + 3*y^2*\sigma^2*\mu + 3*y*\sigma*\mu^2 + \mu^3] \\ &= E[3*y^2*\mu*\sigma^2 + \mu^3] \\ &= \mu^3 + 3*\mu*\sigma^2 \end{split}$$

2.

(a)  $\frac{1}{n} \sum_{i=1}^{n} R_i$  has a normal distribution  $\mathbf{N}(\mu, \sigma^2/n)$ , so

$$E[\hat{\gamma}] - \gamma = \mu^3 + 3 * \frac{\mu}{n} * \sigma^2 - \mu^3 - 3 * \mu * \sigma^2$$
$$= 3 * \frac{(1-n)\mu}{n} * \sigma^2$$

(b) Yes, it's consistent.

$$E[\hat{\gamma}^2] - (E[\hat{\gamma}])^2 = \frac{O(n^6) + 6! * C_6^6 * \sum_{i=1}^6 R_i}{n^6} - \mu^6 - O(\frac{1}{n})$$

$$\xrightarrow{n \to \infty} 0$$

3. The bias of  $(\frac{1}{n}\sum_{i=1}^{n}R_i)^3$  is  $3*\frac{(1-n)\mu}{n}*\sigma^2$ . From previous work, we also have

$$\gamma = \mu^3 + 3 * \mu * \sigma^2$$

$$E[\hat{\gamma}] = \mu^3 + 3 * \frac{\mu}{n} * \sigma^2$$

$$E[R_1R_2R_3] = \mu^3$$

We can easily have an unbiased estimator  $n(\frac{1}{n}\sum_{i=1}^{n}R_i)^3 - (n-1)R_1R_2R_3$ .

4.

(a)

$$E[\tilde{\gamma}] - \gamma = \frac{n}{n} E[R_1^3] - E[R_1^3] - 0$$

(b) Yes, it's consistent.

$$\begin{split} E[\tilde{\gamma}^2] - (E[\tilde{\gamma}])^2 &= \frac{(\sum_{i=1}^n Y_i)^2}{n^2} - (\mu^3 + 3 * \mu * \sigma^2)^2 \\ &= \frac{O(n^2) + n * (n-1)E[R_1^3 R_2^3]}{n^2} - (\mu^3 + 3 * \mu * \sigma^2)^2 \\ &= (\mu^3 + 3 * \mu * \sigma^2)^2 - (\mu^3 + 3 * \mu * \sigma^2)^2 \\ &\stackrel{n \to \infty}{\longrightarrow} 0 \end{split}$$

5. Based on previious work, we have two sufficient statistics  $\bar{R} = \frac{1}{n} \sum_{i=1}^{n} R_i$  and  $\bar{R}^2 = \frac{1}{n} \sum_{i=1}^{n} R_i^2$ . We also have unbiased and consistent estimator  $\tilde{\gamma} = \frac{1}{n} \sum_{i=1}^{n} R_i^3$ . The  $MVUE = E[\tilde{\gamma}|T]$ .

$$E[\tilde{\gamma}|T] = E[\frac{1}{n} \sum_{i=1}^{n} R_{i}^{3} | \bar{R}, \bar{R}^{2}]$$

$$= E[\frac{1}{n} \sum_{i=1}^{n} R_{i} - \bar{R} + \bar{R}^{3} | \bar{R}, \bar{R}^{2}]$$

$$= E[\frac{1}{n} \sum_{i=1}^{n} [(R_{i} - \bar{R})^{3} + 3(R_{i} - \bar{R})^{2} \bar{R} + 3(R_{i} - \bar{R}) \bar{R}^{2} + \bar{R}^{3}] | \bar{R}, \bar{R}^{2}]$$

Similar with the derivation of E(Y) = 0,  $E(Y^3) = 0$ , in question 1, we get

$$E(R_i - \bar{R}) = 0, E((R_i - \bar{R})^3) = 0.$$
 So,

$$E[\tilde{\gamma}|T] = E[\frac{1}{n} \sum_{i=1}^{n} [3(R_i - \bar{R})^2 \bar{R} + \bar{R}^3] | \bar{R}, \bar{R}^2]$$

$$= E[3R_i^2 \bar{R} - 6R_i \bar{R}^2 + 4\bar{R}^3 | \bar{R}, \bar{R}^2]$$

$$= 3\bar{R}_i^2 \bar{R} - 2\bar{R}^3$$