Homework # 9, Stat355, Spring 2021

Please use R to do all of the following problems. Submit your homework by a WORD or PDF document with main results and the corresponding code.

- 1. Confidence intervals are often used in practice. To understand how to implement the method, compute confidence intervals in the following data.
 - (a) Data: $X_1, \dots, X_n \sim^{iid} N(\mu, \sigma^2)$. The z-confidence interval for μ is $\bar{x} \pm z_{\alpha/2} s / \sqrt{n}$. The *t*-confidence interval for μ is $\bar{x} \pm t_{\alpha/2,n-1} s / \sqrt{n}$. Suppose we observe data

5.691, 5.107, 4.644, 4.970, 5.218, 4.418, 5.021, 5.239, 4.752, 6.071.

Compute the 95% z and t confidence intervals for μ , respectively, and compare their lengths. You need to include your mathematical details as well as your R code and output.

Solution; Based on the data, we have $\bar{x} = 5.113$, $s^2 = 0.2379 = 0.4878^2$, n = 10, $z_{0.025} = 1.96$, and $z_{0.025,9} = 2.262$. The 95% z-confidence interval for μ is

$$\bar{x} \pm z_{0.025} \frac{s}{\sqrt{n}} = 5.113 \pm 1.96 \frac{0.4878}{\sqrt{10}} = [4.811, 5.415].$$

The 95% *t*-confidence interval for μ is

$$\bar{x} \pm t_{0.025,9} \frac{s}{\sqrt{n}} = 5.114 \pm 2.262 \frac{0.4878}{\sqrt{10}} = [4.764, 5.462].$$

Their lengths are 0.604 and 0.698, respectively. The *t*-confidence interval is longer. **R code and output**

```
> x <- c(5.691,5.107,4.644,4.970,5.218,4.418,5.021,5.239,4.752,6.071)</pre>
> n <- length(x)
> x.mean <- mean(x)</pre>
> x.var <- var(x)
> x.mean
[1] 5.1131
> x.var
[1] 0.2378894
> qnorm(0.975)
[1] 1.959964
> qt(0.975,n-1)
[1] 2.262157
> c(x.mean-qnorm(0.975)*sqrt(x.var/n), x.mean+qnorm(0.975)*sqrt(x.var/n))
[1] 4.810802 5.415398
> c(x.mean-qt(0.975,n-1)*sqrt(x.var/n),x.mean+qt(0.975,n-1)*sqrt(x.var/n))
[1] 4.764192 5.462008
```

(b) Data: $X \sim Bin(50, \theta)$. The binomial z confidence interval is $\hat{\theta} \pm z_{\alpha/2} \sqrt{\hat{\theta}(1-\hat{\theta})/n}$, where $\hat{\theta} = X/n$. Suppose we observe x = 15. Compute the 95% z confidence interval for θ . You need to include your mathematical details as well as your R code and output. Solution: Based on the data, we have $\hat{\theta} = 15/50 = 0.3$. Thus, the 95% z confidence interval for θ is

$$\hat{\theta} \pm 1.96\sqrt{\frac{\hat{\theta}(1-\hat{\theta})}{n}} = 0.3 \pm 1.96\sqrt{\frac{0.3(1-0.3)}{50}} = [0.1730, 0.4270]$$

R code and output

> n <- 50 > x <- 15 > theta.hat <- x/n > theta.hat [1] 0.3 > c(theta.hat-1.96*sqrt(theta.hat*(1-theta.hat)/n), + theta.hat+1.96*sqrt(theta.hat*(1-theta.hat)/n)) [1] 0.1729775 0.4270225

(c) Data; $X_1, \dots, X_n \sim Poisson(\theta)$. The Poisson z confidence interval is $\bar{x} \pm z_{\alpha/2}\sqrt{\bar{x}/n}$. Suppose we observe data

Compute the 95% z confidence interval for θ . You need to include your mathematical details as well as your R code and output.

Solution: Based on the data, we have $\bar{x} = 20$. Thus, the 95% z-confidence interval for θ is

$$\bar{x} \pm 1.96 \sqrt{\bar{x}/n} = 20 \pm 1.96 \sqrt{20/n} = [16.08, 23.92].$$

> x <- c(21,23,25,12,19)
> x.mean <- mean(x)
> x.mean
[1] 20
> c(x.mean-1.96*sqrt(x.mean/5),x.mean+1.96*sqrt(x.mean/5))
[1] 16.08 23.92

2. z and t confidence intervals are both popular. In particular, let $X_1, \dots, X_n \sim^{iid} N(\mu, \sigma^2)$, where both μ and σ^2 are unknown. Then, the $1 - \alpha$ level z confidence interval for μ is $\bar{x} \pm z_{\alpha/2} s/\sqrt{n}$, and the $1 - \alpha$ level t confidence interval for μ is $\bar{x} \pm t_{\alpha/2,n-1} s/\sqrt{n}$. Use simulation with 10^4 replications to evaluate the coverage probability and the length of the 95% z and t confidence intervals for μ when n = 5, 10, 20, 50, 100 with $\mu = 0, 1, 2, 3, 4, 5$ and fixed $\sigma^2 = 1$, respectively. You need to provide a simple description of your simulation, a summary of your findings, and your R code and output.

Solution: For each combination of μ and n, I generate data 10^4 times. In each replication of the generation, I compute the z and t-confidence intervals for μ . I say the confidence interval

is correct and output 1 if it contains μ use the definition. I also calculate their lengths. I use the average of the correctness of the confidence interval as the approximation of the coverage probability. I use the average of the length as the length of the confidence interval. Al those are derive based on R. I find that the coverage probabilities of z-confidence interval is much lower that 0.95 when n is small, indicating that it is inappropriate for small n (e.g., n = 5). The coverage probabilities of t-confidence interval is always around 0.95, indicating that it is always appropriate. The length of z-confidence interval is much lower that that of t-confidence interval for small n (e.g., n = 5). Their lengths are close to each other when n is not small (e.g., n = 100).

```
> sigma <- 1
> n.o <- c(5,10,20,50,100)
> mu.o <- c(0,1,2,3,4,5)
> run <- 1e4
> COVERAGE <- matrix(0,length(n.o)*length(mu.o),4)</pre>
> LENGTH <- matrix(0,length(n.o)*length(mu.o),4)</pre>
>
> result <- matrix(0,run,4)</pre>
> for(i in 1:length(n.o)){
   for(j in 1:length(mu.o)){
+
    n <- n.o[i]
+
    mu <- mu.o[j]
+
+
    for(k in 1:run){
     x <- rnorm(n,mean=mu,sd=sigma)</pre>
+
     x.mean <- mean(x)</pre>
+
+
     x.var <- var(x)
     CI.z <- c(x.mean-1.96*sqrt(x.var/n), x.mean+1.96*sqrt(x.var/n))
+
     CI.t <- c(x.mean-qt(0.975,n-1)*sqrt(x.var/n), x.mean+qt(0.975,n-1)*sqrt(x.var/n))
+
     result[k,] <- c(CI.z,CI.t)</pre>
+
    }
+
    COVERAGE[length(mu.o)*(i-1)+j,] <- c(n,mu,mean((result[,1]<=mu)</pre>
+
   *(result[,2]>=mu))
+
   ,mean((result[,3]<=mu)*(result[,4]>=mu)))
+
    LENGTH[length(mu.o)*(i-1)+j,] <- c(n,mu,mean(result[,2]-</pre>
+
   result[,1]),mean(result[,4]-result[,3]))
+
   }
+
+ }
> COVERAGE
       [,1] [,2]
                    [,3]
                           [,4]
 [1,]
          5
               0 0.8757 0.9465
 [2,]
          5
               1 0.8772 0.9466
 [3,]
          5
               2 0.8764 0.9507
 [4,]
          5
               3 0.8765 0.9495
               4 0.8772 0.9516
 [5,]
          5
```

[6,]	5	5	0.8769 0.9495
[7,]	10	0	0.9156 0.9450
[8,]	10	1	0.9108 0.9455
[9,]	10	2	0.9150 0.9475
[10,]	10	3	0.9133 0.9467
[11,]	10	4	0.9183 0.9473
[12,]	10	5	0.9183 0.9505
[13,]	20	0	0.9322 0.9495
[14,]	20	1	0.9377 0.9500
[15,]	20	2	0.9383 0.9517
[16,]	20	3	0.9331 0.9484
[17,]	20	4	0.9344 0.9484
[18,]	20	5	0.9334 0.9505
[19,]	50	0	0.9416 0.9473
[20,]	50	1	0.9488 0.9543
[21,]	50	2	0.9458 0.9522
[22,]	50	3	0.9459 0.9523
[23,]	50	4	0.9466 0.9523
[24,]	50	5	0.9424 0.9485
[25,]	100	0	0.9463 0.9490
[26,]	100	1	0.9444 0.9470
[27,]	100	2	0.9509 0.9538
[28,]	100	3	0.9430 0.9468
[29,]	100	4	0.9447 0.9475
[30,]	100	5	0.9470 0.9502
> LENGTH			
	[,1]	[,2]	[,3] [,4]
[1,]	5	0	1.6450092 2.3302438
[2,]	5	1	1.6426055 2.3268388
[3,]	5	2	1.6383204 2.3207687
[4,]	5	3	1.6475883 2.3338972
[5,]	5	4	1.6428533 2.3271898
[6,]	5	5	1.6466438 2.3325592
[7,]	10	0	1.2061673 1.3921122
[8,]	10	1	1.2051346 1.3909204
[9,]	10	2	1.2011886 1.3863660
[10,]	10	3	1.2057224 1.3915988
[11,]	10	4	1.2071103 1.3932006
[12,]	10	5	1.2036975 1.3892616
[13,]	20	0	0.8661867 0.9249743
[14,]	20	1	0.8637670 0.9223904
[15,]	20	2	0.8672357 0.9260945
[16,]	20	3	0.8647519 0.9234421

[17,]	20	4	0.8632113	0.9217969
[18,]	20	5	0.8679184	0.9268235
[19,]	50	0	0.5520224	0.5659849
[20,]	50	1	0.5527567	0.5667379
[21,]	50	2	0.5518809	0.5658399
[22,]	50	3	0.5519585	0.5659195
[23,]	50	4	0.5522228	0.5661904
[24,]	50	5	0.5510085	0.5649454
[25,]	100	0	0.3916174	0.3964561
[26,]	100	1	0.3911581	0.3959911
[27,]	100	2	0.3912275	0.3960613
[28,]	100	3	0.3912050	0.3960386
[29,]	100	4	0.3911971	0.3960306
[30,]	100	5	0.3910814	0.3959134

3. z confidence interval is often used in binomial data for success/failure experiments. In particular, suppose the success/failure experiment is repeated n times, such that the total number of successes $X \sim Bin(n,\theta)$, where θ is the probability of success in each experiment. Then, the $1 - \alpha$ level z confidence interval for θ is $\hat{\theta} \pm z_{\alpha/2}\sqrt{\hat{\theta}(1-\hat{\theta})/n}$, where $\hat{\theta} = X/n$. Use simulation with 10^4 replications to evaluate the coverage probability of the 95% z confidence interval for θ when θ is close 0, where you choose $\theta = 0.01, 0.02, 0.05, 0.1$ with n = 10, 20, 50, 100, 200, 500, 1000, respectively. You need to provide a simple description of your simulation, a summary of your findings, and your R code and output.

Solution: For each combination of n and θ with n = 10, 20, 50, 100, 200, 500, 1000 and $\theta = 0.01, 0.02, 0.05, 0.1$, I generate data from $Bin(n, \theta)$ with 10^4 replications. In each replication, I calculate the 95% z-confidence interval for θ . I denote 1 if the confidence interval contains the true value of θ or 0 otherwise. The coverage probability is approximated by the proportion of 1. The result shows that the coverage probability is much lower than 0.95 for small n (e.g., n = 10, 20, 50, 100) when θ is very close to 0 (e.g., $\theta = 0.01, 0.02$). The coverage probability increases as either n or θ increases. I reach 0.95 when n is large (e.g. n = 1000) in all the cases of θ we studied.

```
> n.o <- c(10,20,50,100,200,500,1000)
> theta.o <- c(0.01,0.02,0.05,0.1)
> run <- 1e4
> COVERAGE <- matrix(0,length(n.o)*length(theta.o),3)
> for(i in 1:length(n.o)){
+ for(j in 1:length(theta.o)){
+     n <- n.o[i]
+     theta <- theta.o[j]
+     x <- rbinom(run,n,theta)
+     theta.hat <- x/n
+     lower <- theta.hat-1.96*sqrt(theta.hat*(1-theta.hat)/n)</pre>
```

```
+
   upper <- theta.hat+1.96*sqrt(theta.hat*(1-theta.hat)/n)</pre>
   COVERAGE[(i-1)*length(theta.o)+j,] <- c(n,theta,mean((lower<=theta)*(upper>=theta)))
+
+ }
+ }
> dimnames(COVERAGE)[[2]] <- c("n","theta","coverage")</pre>
> COVERAGE
        n theta coverage
 [1,]
        10 0.01
                  0.0964
 [2,]
        10 0.02
                  0.1807
 [3,]
        10 0.05
                  0.3994
 [4,]
        10 0.10
                  0.6461
 [5,]
        20 0.01
                  0.1825
 [6,]
        20 0.02
                  0.3363
 [7,]
        20 0.05
                  0.6333
 [8,]
        20 0.10
                  0.8756
        50 0.01
 [9,]
                  0.3969
[10,]
        50 0.02
                  0.6374
[11,]
       50 0.05
                  0.9229
[12,]
       50 0.10
                  0.8790
[13,]
      100 0.01
                  0.6443
[14,]
      100 0.02
                  0.8658
      100 0.05
[15,]
                  0.8823
[16,]
      100 0.10
                  0.9301
[17,]
      200 0.01
                  0.8654
[18,]
      200 0.02
                  0.9025
[19,]
      200 0.05
                  0.9264
      200 0.10
[20,]
                  0.9282
[21,]
      500 0.01
                  0.8724
[22,]
      500 0.02
                  0.9274
[23,] 500 0.05
                  0.9310
[24,] 500 0.10
                  0.9397
[25,] 1000 0.01
                  0.9263
[26,] 1000 0.02
                  0.9504
[27,] 1000 0.05
                  0.9468
```

[28,] 1000 0.10

0.9537