STAT430 Homework #5: Due Friday, March 8, 2019.

Name:

- 0. We will continue with **Chapter 8** on estimation. You should have read **Sections 8.1-8.4** on point estimation; please continue with **Sections 8.5-8.10** on interval estimation and related topics.
- 1. Suppose $[\hat{\theta}_L, \hat{\theta}_U]$ is a $(1-\alpha)$ 100% confidence interval for the target parameter θ .
- (a) Find a (1α) 100% confidence interval for the new target parameter $a + b\theta$, where a is a known real number and b > 0 is a known real number.

Answer:

$$1 - \alpha = P(\hat{\theta}_L \le a + b\theta \le \hat{\theta}_U), b > 0$$

$$= P(\frac{\hat{\theta}_L - a}{b} \le \theta \le \frac{\hat{\theta}_U - a}{b})$$

$$[\frac{\hat{\theta}_L - a}{b}, \frac{\hat{\theta}_U - a}{b}]$$

(b) Find a $(1 - \alpha)$ 100% confidence interval for the new target parameter $a + b\theta$, where a is a known real number and b < 0 is a known real number.

Answer:

$$\begin{split} 1 - \alpha &= P(\hat{\theta}_L \leq a + b\theta \leq \hat{\theta}_U), b < 0 \\ &= P(\frac{\hat{\theta}_L - a}{b} \geq \theta \geq \frac{\hat{\theta}_U - a}{b}) \\ &[\frac{\hat{\theta}_U - a}{b}, \frac{\hat{\theta}_L - a}{b}] \end{split}$$

(c). Suppose Y_1, \ldots, Y_{16} are iid $N(\mu, \sigma^2)$ and are measured in degrees Fahrenheit. Form a 90% confidence interval for the true mean μ if the realized data are as follows:

 $Y \leftarrow c(97.91, 98.81, 97.42, 98.01, 98.25, 98.18, 98.90, 98.25, 99.59, 98.50, 98.60, 97.92, 97.48, 99.21$

Answer:

This is the generic 90% CI for \bar{Y}

$$\bar{Y} \pm t_{\alpha/2,n-1} \sqrt{S^2/n}$$

Y <- c(97.91, 98.81, 97.42, 98.01, 98.25, 98.18, 98.90, 98.25, 99.59, 98.50, 98.60, 97.92, 97.48, 99.21

Y_bar <- mean(Y)

lower <- Y_bar + qt(.05,length(Y))*sqrt(var(Y)/length(Y))

```
upper <- Y_bar - qt(.05,length(Y))*sqrt(var(Y)/length(Y))</pre>
lower
```

[1] 98.15985

upper

[1] 98.6914

So a 90% CI for \bar{Y} is,

[98.15985, 98.6914]

(d). Suppose we are interested in the true mean in degrees Celsius = (5/9) * (Fahrenheit - 32), instead of Fahrenheit. Use the data in part (c) to compute a 90% confidence interval for the true mean in degrees Celsius.

Answer:

```
X < -c(97.91, 98.81, 97.42, 98.01, 98.25, 98.18, 98.90, 98.25, 99.59, 98.50, 98.60, 97.92, 97.48, 99.21
Y \leftarrow (5/9)*(X -32)
Y bar <- mean(Y)
lower <- Y_bar + qt(.05,length(Y))*sqrt(var(Y)/length(Y))</pre>
upper <- Y_bar - qt(.05,length(Y))*sqrt(var(Y)/length(Y))
lower
```

[1] 36.75547

upper

[1] 37.05078

So a 90% CI for the temperature in Celsius is,

[36.75547, 37.05078]

2. The code in the answer block below computes and plots exact $(1-\alpha)$ 100% two-sided confidence intervals for nreps replicates of the sampling experiment in which Y_1, \ldots, Y_n iid $N(\theta, \sigma^2)$, with θ as our target and σ^2 unknown. The pivot used to produce the CI is

$$Q = \frac{\bar{Y} - \theta}{\sqrt{S^2/n}} \sim t_{n-1},$$

where \bar{Y} is the sample mean of the observations and S^2 is the sample variance. The quantiles then come from the t distribution with n-1 degrees of freedom, and by isolating θ in the probability statement, we get

$$1 - \alpha = P\left(q_{\alpha/2} \le Q \le q_{1-\alpha/2}\right) = P\left(\bar{Y} + q_{\alpha/2}\sqrt{\frac{S^2}{n}} \le \theta \le \bar{Y} + q_{1-\alpha/2}\sqrt{\frac{S^2}{n}}\right),$$

so

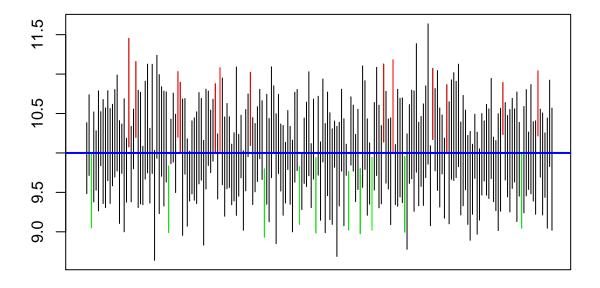
$$\hat{\theta}_L = \bar{Y} + q_{\alpha/2} \sqrt{\frac{S^2}{n}}, \quad \hat{\theta}_U = \bar{Y} + q_{1-\alpha/2} \sqrt{\frac{S^2}{n}}.$$

(Because the t distribution is symmetric about zero, $q_{\alpha/2} = -q_{1-\alpha/2}$, and we often write $\bar{Y} \pm q_{1-\alpha/2}\sqrt{S^2/n}$ for the CI.)

Run the code below with the given settings, then increase the sample size to n = 100 and run again. Comment on the width and the *coverage* (empirical proportion of hits) of the CIs at the two sample sizes. What changes and what does not change?

Answer:

```
nreps <- 200
theta <- 10
n <- 10
alpha <- 0.1
set.seed(4302019)
Y <- rnorm(n * nreps, mean = theta, sd = 1)
YY <- matrix(Y, n, nreps) # nreps iid samples of size n, one in each column
sample_mean <- apply(YY, MAR = 2, FUN = "mean") # sample mean, Ybar, for each column</pre>
sample_var <- apply(YY, MAR = 2, FUN = "var") # sample variance, S^2, for each column
# Compute lower confidence limit, using quantile from t distribution
theta_hat_L <- sample_mean + qt(alpha / 2, df = n - 1) * sqrt(sample_var / n)
# Compute upper confidence limitk using quantile from t distribution
theta_hat_U <- sample_mean + qt(1 - alpha / 2, df = n - 1) * sqrt(sample_var / n)
too_high <- (theta < theta_hat_L) # CI is too high and missed theta
too_low <- (theta > theta_hat_U) # CI is too low and missed theta
mean(too_low + too_high)
                          # proportion of misses should be close to alpha
## [1] 0.115
1 - mean(too_low + too_high) # proportion of hits should be close to 1 - alpha
## [1] 0.885
# Plot the CIs.
plot(c(0,1), c(min(theta_hat_L), max(theta_hat_U)), type = "n",
     xlab = paste(nreps, "replicates of confidence intervals"), ylab = "", xaxt = "n")
x <- (1:nreps) / nreps # index for each of the nreps CIs
segments(x, theta_hat_L, x, theta_hat_U)
segments(x[too_high], theta_hat_L[too_high], x[too_high], theta_hat_U[too_high], col = "red")
segments(x[too_low], theta_hat_L[too_low], x[too_low], theta_hat_U[too_low], col = "green")
abline(h = theta, col = "blue", lwd = 2)
```



200 replicates of confidence intervals

3. In class, we considered the sampling experiment in which Y_1, \ldots, Y_n iid $N(\mu, \sigma^2)$, with $\theta = \sigma^2$ as our target. The pivot used to produce the CI is

$$Q = \frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2,$$

where S^2 is the sample variance. The quantiles then come from the χ^2 distribution with n-1 degrees of freedom, and by isolating θ in the probability statement, we obtained the two-sided CI,

$$\left[\frac{(n-1)S^2}{q_{1-\alpha/2}}, \frac{(n-1)S^2}{q_{\alpha/2}}\right].$$

Modify the code from problem 2 above to compute and plot nreps=200 95% confidence intervals for σ^2 when n=20 and the true population parameters are $\mu=10$, $\sigma^2=1$. Report your empirical coverage (proportion of hits).

Answer:

- 4. Consider the setting of problem 3. The true variance, σ^2 , is naturally bounded below by zero. In some cases, we might only be interested in finding an upper confidence limit for σ^2 , so that we have the one-sided confidence interval $(-\infty, \hat{\theta}_U]$, but we know this is actually $[0, \hat{\theta}_U]$. For this problem, the target is still σ^2 and we still use the same pivot.
- (a). Show that

$$1 - \alpha = P\left(0 \le \sigma^2 \le \frac{(n-1)S^2}{q_\alpha}\right),\,$$

so that

$$\left[0, \frac{(n-1)S^2}{q_\alpha}\right]$$

is a valid $(1 - \alpha)$ 100% confidence interval for σ^2 .

Answer:		
. ,	e setting as problem 3, modify the code to compute a port their empirical coverage. In this case, you can rep_L <- rep(0, nreps).	_
Answer:		
5. Complete Exercise 8	$\bf 8.40$ of the text. In this problem, the variance is $know$	n to be 1.
Answer:		
6, parts 6(a) and 6(b). Con	mplete parts (a) and (b) of Exercise 8.44 of the text.	
Answer:		
6(c). Quantiles are defined	by the property that	
	$F_Q(q_\alpha) = P(Q \le q_\alpha) = \alpha,$	
	tive distribution function of the random variable, Q . Use the pression for the α th quantile of the pivotal distribution, $q_{0.9}, q_{0.95}$, and $q_{0.975}$.	
Answer:		
6(d). Use your computed qu	uantiles and pivotal quantity to compute a 90% two-side	ed confidence interval for θ .
Answer:		
	quantiles and pivotal quantity to compute the 90% up of the book, so you can check your answer in the back	
Answer:		
	inplete Exercise 8.47 of the text. (d) Then repeat part with $n = 21$. Comment on the difference between the	

Answer: