

Textbook Exercises

6.12, 6.20, 6.38, 6.50, 6.64, 6.70, 6.84, 6.97, 6.120, 6.130, 6.145, 6.150

Computer Exercises

For each R problem, turn in answers to questions with the written portion of the homework. Send the R code for the problem to Katherine Goode. The answers to questions in the written part should be well written, clear, and organized. The R code should be commented and well formatted.

R problem 1 Ideally, a 95% confidence interval will be as tightly clustered around the true value as possible, and will have a 95% coverage probability. When the possible data values are discrete, (such as in the case of sample proportions which can only be a count over the sample size), the true coverage or capture probability is not exactly 0.95 for every p . This problem examines the true coverage probability for three different methods of making confidence intervals.

To compute the coverage probability of a method, recognize that each possible value x from 0 to n for a given method results in a confidence interval with a lower bound $a(x)$ and an upper bound $b(x)$. The interval will capture p if $a(x) < p < b(x)$. To compute the capture probability of a given p , we need to add up all of the binomial probabilities for the x values that capture p in the interval. For a sample size n and true population proportion p , this coverage probability is

$$P(p \text{ in interval}) = \sum_{x:a(x) < p < b(x)} \binom{n}{x} p^x (1-p)^{n-x}$$

To compute this in R, you need to find the lower and upper bounds of the confidence interval for each possible outcome x , and add the probabilities of the outcomes that capture p . Here is some code to get you started using the textbook method for an example where $x = 10$ and $p = 0.3$.

```
x = 0:10
p.hat = x/10 # will be a vector
se = sqrt(p.hat*(1-p.hat)/10) # also a vector
z = qnorm(0.975)
a = p.hat - z*se # also a vector
b = p.hat + z*se #also a vector
x[ (a < 0.3) & (0.3 < b) ] # x that capture p

## [1] 2 3 4 5 6
```

For each of the following methods, find which outcomes x result in confidence intervals that capture p and compute the coverage probability from a sample of size $n = 60$ when $p = 0.4$.

1. Normal from maximum likelihood estimate, $\hat{p} = X/n$, $SE = \sqrt{\hat{p}(1 - \hat{p})/n}$, with the interval

$$\hat{p} \pm 1.96SE$$

2. Normal from adjusted maximum likelihood estimate, $\tilde{p} = (X+2)/(n+4)$, $SE = \sqrt{\tilde{p}(1 - \tilde{p})/(n + 4)}$, with the interval

$$\tilde{p} \pm 1.96SE$$

3. Within $z^2/2$ of the maximum likelihood loglikelihood. For this method, the file `log1.R` has a function `log1.ci.p()` which returns the lower and upper bounds of a 95% confidence interval given n and x . You can graph the loglikelihood using `glog1.p()` for n , x , and $z = 1.96$ to see if the returned values make sense.

R Problem 2 Repeat the previous problem, but for $n = 60$ and $p = 0.1$.

R Problem 3 This problem examines a t distribution with 4 degrees of freedom.

1. Draw a graph of a t distribution with 4 degrees of freedom and a standard normal curve from -4 to 4 .
2. Find the area to the right of 2 under each curve.
3. Find the 0.975 quantile of each curve.

R Problem 4 Repeat the previous problem, but for a t distribution with 20 degrees of freedom.

R Problem 5 Repeat the previous problem, but for a t distribution with 100 degrees of freedom.

Here is some sample code to draw graphs of continuous distributions.

```
x = seq(-4,4,0.001)
z = dnorm(x)
y.10 = dt(x, df=10)
d = data.frame(x,z,y.10)
require(ggplot2)

## Loading required package: ggplot2

ggplot(d) +
  geom_line(aes(x=x,y=y.10),color="blue") +
  geom_line(aes(x=x,y=z),color="red") +
  ylab('density') +
  ggtitle("t(10) distribution in blue, N(0,1) in red")
```

