# For Loops

## Contents

Simulating the Law of Large Numbers	1
Extending Our Simulation	2

3

### Making the code more efficient.

### Simulating the Law of Large Numbers

The Law of Large Numbers says that as sample sizes increase, the mean of the sample will approach the true mean of the distribution. We are going to simulate this phenomenon!

We'll start by making a vector of sample sizes from 1 to 50, to represent increasing sample sizes.

Create a vector called sample\_sizes that is made up of the numbers 1 through 50. (Hint: You can use seq() or : notation).

We'll make an empty tibble to store the results of the for loop:

estimates <- tibble(n = integer(), sample\_mean = double())</pre>

Write a loop over the sample\_sizes you specified above. In the loop, for each sample size you will:

- 1. Calculate the mean of a sample from the random normal distribution with mean = 0 and sd = 5.
- 2. Make an intermediate tibble to store the results
- 3. Append the intermediate tibble to your tibble using bind\_rows().

```
set.seed(60637)
for (___ in ___) {
    # Calculate the mean of a sample from the random normal distribution with mean = 0 and sd = 5.
    ___ <- ___
    # Make a tibble with your estimates
    this_estimate <- tibble(n = ___, sample_mean = ___)
    # Append the new rows to your tibble
    ___ <- bind_rows(estimates, ___)
}</pre>
```

We can use ggplot2 to view the results. Fill in the correct information for the data and x and y variables, so that the n column of the estimates tibble is plotted on the x-axis, while the sample\_mean column of the estimates tibble is plotted on the y-axis.

```
# your data goes in the first position
____%>%
ggplot(aes(x = ___, y = ___)) +
geom line()
```

1. As the sample size (n) increases, does the sample mean becomes closer to 0, or farther away from 0?

Rewrite the loop code without looking at your previous code and use a wider range of sample sizes. Try several different sample size combinations. What happens when you increase the sample size to 100? 500? 1000? Use the seq() function to generate a sensibly spaced sequence.

```
set.seed(60637)
sample_sizes <- ____
estimates_larger_n <- ____
for (___ in ___) {
    ___ <- ___
    ___ <- ___
    ___ <- ___
}
___ %>%
ggplot(___(__ = ___, ___ = ___)) +
    geom_line()
```

1. How does this compare to before?

#### **Extending Our Simulation**

Looking at your results, you might think a small sample size is sufficient for estimating a mean, but your data had a relatively small standard deviation compared to the mean. Let's run the same simulation as before with different standard deviations.

Do the following:

- 1. Create a vector called population\_sd of length 4 with values 1, 5, 10, and 20 (you're welcome to add larger numbers if you wish).
- 2. Make an empty tibble to store the output. Compared to before, this has an extra column for the changing population standard deviations.
- 3. Write a loop inside a loop over population\_sd and then sample\_sizes.
- 4. Then, make a ggplot graph where the x and y axes are the same, but we facet (aka we create small multiples of individual graphs) on population\_sd.

```
set.seed(60637)
population_sd <- ____
# use what every you came up with in the previous part
sample_sizes <- ____
estimates_adjust_sd <- ____
for (____ in ____) {
    for (____ in ____) {
        ____ <- ___
        ____ <- ___
        ____ <- ____
}
}</pre>
```

```
___ %>%
ggplot(___) +
geom_line() +
facet_wrap(~population_sd) +
theme_minimal()
```

How do these estimates differ as you increase the standard deviation?

## Making the code more efficient.

At this point in your coding career, efficiency is minimally important. However, we discussed that loops in R are much slower if we build our output item by item (or row by row in this case.) We do better if we preallocate space.

- 1. Rewrite your loop from part 1 where we start with estimates <- vector("list", n). This creates a list of length n "filled" with NULL.
- 2. During each iteration of your loop fill one of the spaces in estimates with a named vector (which we think of as a row.) Use [[ to index into your list.
- 3. Finally, when your loop terminates, you'll have estimates filled with n "rows". bind\_rows(estimates) will convert the list to a tibble.

Want to improve this tutorial? Report any suggestions/bugs/improvements on here! We're interested in learning from you how we can make this tutorial better.