

# Report

## Overview

The goal of this project was to assign probabilities to the upcoming Georgia Senate Election Runoff between David Perdue and Jon Ossoff. I approached this project with a combination of prior political science understanding as well as an implementation of simulation techniques. The end result of this project is a graphic that shows the user the projected probabilities for each candidate. Additionally this simulation can be updated as new polls come in.

## Implementation

Every file within this repo is self contained and will run independently. Within the repo is a package called `georgia` which holds a few of the key functions used in the project. The three functions it contains are `probability_winning_plot`, `probability_dataset`, and `initial_poll_cleaning`. The goal of these three functions is to take dataset from `fivethirtyeight.com` and allow them to be easily implemented in the probability forecast model. `initial_poll_cleaning` takes in a downloaded `.csv` file that is saved as a variable and does the data manipulation necessary to proceed with the forecasting. The function `probability_dataset` works on this by doing some of the more complicated data manipulation, involving logical statements, and make it ready for the last function, `probability_winning_plot`. `probability_winning_plot` then plots the probabilities in a way that allows for dynamic poll values.

## Bootstrapping and Mapping

In addition to the functions described above, there are the actual simulation techniques used to come up with the probabilities. At the time of writing there are 114 polls for this particular Senate race. Although this is actually a decent amount of polling compared to most senate races, in my opinion it was not enough to make an accurate prediction. My first attempt to model the uncertainty in the election can be seen in my file `01_Johnson_Distribution.Rmd`. Although this did not pass the mathematical muster required of an election simulation it was an important part of the process towards finding a good way to go about the project. It led to the implementation of bootstrapping, which can be found in `02_Bootstrap.Rmd`. After cleaning the polling dataset, resulting in the named dataset `boot_data`, I performed the following bootstrap:

```
map(1:10000, ~sample(boot_data$actual_spread, size = length(boot_data), replace = TRUE))
%>% map_dbl(mean)
```

This line of code takes samples from the `boot_data` with replacement and then maps over them ten thousand times and lastly extracting the mean. This allowed me to achieve normality and make some predictions.

## Assigning Probability Values