# Multinomial CI Prediction for 2020 Election

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2020-11-06

```
Load package
```

```
library(jsonlite)
library(lubridate)
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
       date, intersect, setdiff, union
library(MultinomialCI)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
##
       intersect, setdiff, setequal, union
You also need to load the data. You should have the all.state.dataset.
load("MultinomialCIProject2020.RData")
all.state.dataset <- get('all.state.dataset')</pre>
```

#### Predifine function

The script concerns about the probability of a vote cast for a specific candidate. Assuming the votes follow a multinomial distribution, it uses the MultinomialCI package to calculate the confidence interval of the probability for each candidate.

```
CI.by.county <- function(df, alpha) {
    df <- filter(df, df[[1]] > 0)
    candidate.names <- names(df)
    county <- row.names(df)
    mat <- t(apply(df, 1, as.numeric))

CI.per.county <- function(vec) {
        CI.per.county <- c(multinomialCI(pmax(vec, 0), alpha))
    }
    cols <- apply(expand.grid(candidate.names, c('low', 'high')), 1, paste, collapse='.')
    all.ci <- t(apply(mat, 1, CI.per.county))

    output <- data.frame(all.ci, row.names = county)
    colnames(output) <- cols

    return(output)
}</pre>
```

With the probability of a vote cast for a candidate, the final projection is done by  $np \pm z \cdot \sqrt{n \cdot p \cdot (1-p)}$ , where p is the probability of a candidate winning a vote, n is the expected remaining vote according to NYT.

```
#' Calculate the lower and the upper bound of the remaining votes given the
#' probability.
#'
#' Cparam remaining A scalar or vector containing the remaining votes in the
#1
                    county.
#' Cparam prob A scalar or vector containing the probability of votes going to
#'
               the candidate in the county.
#' @param alpha confidence level
# '
#' @return a length 2 vector, first is the lower bound and the second is the
#'
           upper.
#' @export
range_est <- function(remaining, prob, alpha=0.95) {</pre>
 lo <- remaining*prob + qnorm((1-alpha)/2)*sqrt(remaining*prob*(1-prob))</pre>
 hi <- remaining*prob - qnorm((1-alpha)/2)*sqrt(remaining*prob*(1-prob))
 return(c(sum(lo, na.rm = TRUE), sum(hi, na.rm = TRUE)))
}
#' Helper function to convert UTC time string to EST
#'
#' Oparam timestring string representation of time, NYT's time in ISO format
# '
#' @return a POSIXct object in EST
#' @export
#' @examples to_EST('2020-11-07T01:46:10Z')
to.EST <- function(timestring) {</pre>
  with_tz(parse_date_time(timestring, 'ymd HMS'), 'EST')
}
```

## Pull in data from NYT's api

Need to change the current.state.name — state name used to pull data from NYT's api.

It should be spelled out and all lower case and white space should be replace by -. e.g.

- New York: new-york
- Pennsylvania: pennsylvania
- Distinct of Columbia: district-of-columbia

It will store all the pulled data in all.state.dataset. Only new data will be stored.

```
# This script will load data for the following state.
current.state.name <- 'arizona'</pre>
nyt.api <- paste(</pre>
  'https://static01.nyt.com/elections-assets/2020/data/api/2020-11-03/race-page/',
  current.state.name,
  '/president.json',
  sep = '')
results <- from JSON (nyt.api)
current.update.time <- max(results$data$races$counties[[1]]$last_updated)</pre>
# if the state haven't been track yet, create a slot for that.
if (!current.state.name %in% names(all.state.dataset)) {
  update.type <- "New data"
  all.state.dataset[[current.state.name]] <- list()</pre>
 all.state.dataset[[current.state.name]][[current.update.time]] <- results$data$races$counties[[1]]
  previous.update.time <- 'N/A'
} else { # if the state has been track, see if this one is new update
  previous.update.time <- max(last(all.state.dataset[[current.state.name]])</pre>
                               $last_updated)
  if (current.update.time == previous.update.time) {
    update.type <- "No update"</pre>
  } else {
    update.type <- "New update"</pre>
    all.state.dataset[[current.state.name]][[current.update.time]] <- results$data$races$counties[[1]]
}
rm(results)
by.county <- last(all.state.dataset[[current.state.name]])</pre>
old.by.county <- nth(all.state.dataset[[current.state.name]],</pre>
                      max(1, length(all.state.dataset[[current.state.name]])-1))
all.votes <- data.frame(by.county$results,
                         row.names = by.county$name)
old.all.votes <- data.frame(old.by.county$results,</pre>
                             row.names = old.by.county$name)
mail.votes <- data.frame(by.county$results_absentee,
                          row.names = by.county$name)
old.mail.votes <- data.frame(old.by.county$results_absentee,</pre>
                              row.names = old.by.county$name)
```

```
cat(paste(update.type, 'in', current.state.name),
    paste("Previous Update:",
          to.EST(max(old.by.county$last_updated))),
    paste("Current Update:",
          to.EST(max(by.county$last_updated))),
    paste("Previous margin:",
          sum(old.all.votes$bidenj) - sum(old.all.votes$trumpd)),
    paste("Current margin:",
          sum(all.votes$bidenj) - sum(all.votes$trumpd)),
    sep='\n')
## No update in arizona
## Previous Update: 2020-11-06 21:01:47
## Current Update: 2020-11-06 21:04:35
## Previous margin: 29861
## Current margin: 29861
See the number of snapshots saved
sapply(all.state.dataset, length)
##
        arizona pennsylvania
                                                               alaska
                                    nevada
                                                georgia
##
                                                                    1
See the time at which snapshots were taken (in EST)
to.EST(names(all.state.dataset[[current.state.name]]))
## [1] "2020-11-06 11:09:33 EST" "2020-11-06 15:27:34 EST"
## [3] "2020-11-06 20:25:37 EST" "2020-11-06 21:01:47 EST"
## [5] "2020-11-06 21:04:35 EST"
```

## Estimate the probability for each candidate

You can change using to update how you want to estimate the probability. Unhide one of them and hide to other to use.

first option:

- Use the difference between old data and new data
- Best for predicting the most recent trend
- Doesn't work if the difference between old data and new data is small or non-representative

#### second option:

- Use the mail.votes to predict old data and new data
- Work the best if the mail votes is homogeneous throughout different time
- Doesn't work But the demographics within mail data can change over time

When using is incomplete, the probability of each candidate will based on all.votes

```
# first option
using <- data.frame(
  data.matrix(last(all.state.dataset[[current.state.name]])$results)
  - data.matrix(nth(all.state.dataset[[current.state.name]], 1)$results),
  row.names = last(all.state.dataset[[current.state.name]])$name)
# second option</pre>
```

```
# using <- mail.votes
remaining <- data.frame(exp.remaining = pmax(0, by.county$tot_exp_vote - rowSums(data.matrix(all.votes)
                        row.names = by.county$name)
ci.mail <- merge(remaining, CI.by.county(using, 0.95), by=0)
ci.other <- merge(remaining, CI.by.county(all.votes, 0.95), by=0)</pre>
ci.other <- ci.other[ci.other$Row.names %in%</pre>
                       setdiff(ci.other$Row.names, ci.mail$Row.names), ]
est <- rbind(ci.mail, ci.other)
rm(remaining)
est <- est[order(est$Row.names), ]</pre>
rownames(est) <- est$Row.names</pre>
est$Row.names <- NULL
est[est$exp.remaining > 0, ]
##
              exp.remaining bidenj.low trumpd.low jorgensenj.low write.ins.low
## Apache
                       9667
                             0.6807217
                                        0.3074426
                                                       0.01060288 0.0000000000
## Cochise
                       6619
                             0.3402394
                                        0.6233886
                                                       0.03222836 0.0000000000
## Coconino
                        890 0.4689542 0.5000000
                                                       0.02614379 0.0000000000
## La Paz
                        323
                             0.3062753 0.6794968
                                                       0.01108282 0.0001497679
## Maricopa
                      91437
                             0.3953649
                                        0.4829649
                                                       0.02604698 0.0934567668
## Mohave
                       2164
                             0.1743745 0.8003538
                                                       0.02223907 0.0000000000
## Navajo
                       5536 0.3179027 0.6565895
                                                       0.02125650 0.0000000000
## Pima
                      25936 0.4366584 0.4743104
                                                       0.02810825 0.0546476664
## Pinal
                      33500
                             0.4015902
                                        0.5793579
                                                       0.01320132 0.0022502250
## Santa Cruz
                       1454 0.6707766
                                       0.3156144
                                                       0.01089737 0.0006650977
## Yuma
                       3855 0.4859678 0.4810153
                                                       0.02930252 0.0000000000
##
              bidenj.high trumpd.high jorgensenj.high write.ins.high
                0.6816159
                            0.3083368
                                            0.01149715
                                                         0.0004832989
## Apache
## Cochise
                0.3438750
                            0.6270242
                                           0.03586394
                                                         0.0022543675
## Coconino
                0.4747368
                            0.5057826
                                           0.03192639
                                                         0.0041486084
## La Paz
                0.3080035
                            0.6812250
                                           0.01281100
                                                         0.0018779504
## Maricopa
                0.3964692
                            0.4840692
                                           0.02715123
                                                         0.0945610212
## Mohave
                0.1766075
                            0.8025868
                                           0.02447208
                                                         0.0012221448
## Navajo
                0.3211672
                            0.6598541
                                           0.02452105
                                                         0.0018474520
## Pima
                0.4399260
                            0.4775780
                                           0.03137591
                                                         0.0579153239
## Pinal
                0.4036230
                                           0.01523417
                                                         0.0042830783
                            0.5813908
## Santa Cruz
                0.6718347
                            0.3166725
                                           0.01195543
                                                         0.0017231596
## Yuma
                0.4885116
                            0.4835590
                                           0.03184627
                                                         0.0013056215
```

#### Estimate the final range

With the lower and upper bound of Biden's and Trump's probability in a county, we use both probability to calculate the CI. The lower end of the CI from the low probability is a candidate's lower bound, while the upper end of the CI from the high probility is a candidate's upper bound. We then see the margin using Biden's lower bound — Trump's upper bound, and using Biden's upper bound — Trump's lower bound, to calculate the final projection interval.

```
future.lo <- range_est(est$exp.remaining, est$bidenj.low)[1] - range_est(est$exp.remaining, est$trumpd.future.hi <- range_est(est$exp.remaining, est$bidenj.high)[2] - range_est(est$exp.remaining, est$trumpd current.diff <- sum(all.votes$bidenj) - sum(all.votes$trumpd)
current.diff + c(future.lo, future.hi)</pre>
```