

Analysing Twitter for Ubisoft

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8.1 Analysing the Relationship Between Friends and Followers for Twitter Users

8.1.1 Retrieve the posts from Twitter

relevant posts can be retrieved from twitter by utilising the `rtweet` package, packages can be loaded for use in **R** thusly:

The `rtweet` API will search for tweets that contain all the words of a query regardless of uppercase or lowercase usage [kearney2019].

In order to leverage the *Twitter* API it is necessary to use tokens provided through a *Twitter* developer account:

and hence all tweets containing a mention of *Ubisoft* can be returned and saved to disk as shown in listing 3:

8.2.2 Count of Followers and Friends

In order to identify the number of users that are contained in the *tweets* the `unique()` function can be used to return a vector of names which can then be passed as an index to the vector of counts as shown in listing 4, this provides that 81.7% of the tweets are by unique users.

```

1  # Load Packages
   ↪ -----
2  setwd("~/Dropbox/Notes/DataSci/Social_Web_Analytics/SWA-Project/scripts")
   ↪ "/"
3
4  if (require("pacman")) {
5    library(pacman)
6  } else{
7    install.packages("pacman")
8    library(pacman)
9  }
10
11 pacman::p_load(xts, sp, gstat, ggplot2, rmarkdown, reshape2,
12               ggmap, parallel, dplyr, plotly, tidyverse,
13               reticulate, UsingR, Rmpfr, swirl, corrplot,
14               gridExtra, mise, latex2exp, tree, rpart,
15               lattice, coin, primes, epitools, maps, clipr,
16               ggmap, twitterR, ROAuth, tm, rtweet, base64enc,
17               httpuv, SnowballC, RColorBrewer, wordcloud,
18               ggwordcloud, tidyverse, boot)

```

Listing 1: Load the Packages for *R*

8.1.3 Summary Statistics

The average number of friends and followers from users who posted tweets mentioning *Ubisoft* can be returned using the `mean()` as shown in listing 5 this provides that on average each user has 586 friends and 63,620 followers.

8.1.4 Above Average Followers

Each user can be compared to the average number of followers, by using a logical operator on the vector (e.g. `y > ybar`), this will return an output of logical values. *R* will coerce logicals into 1/0 values meaning that the mean value will return the proportion of TRUE responses as shown in listing 6. This provides that 20.6% of the users identified have above average friend counts, while only 2.4% have an above average number of followers.

8.1.5 Generate a bootstrap distribution of the Follower Counts

A bootstrap assumes that the population is an infinitely large repetition of the sample, a bootstrap of the follower counts can be produced by resampling with replacement/repetition and plotted using the `ggplot2` library as shown in listing 7 and figure 1.

```

1  # Set up Tokens
   ↪ =====
2
3  options(RCurlOptions = list(
4    verbose = FALSE,
5    capath = system.file("CurlSSL", "cacert.pem", package = "RCurl"),
6    ssl.verifypeer = FALSE
7  ))
8
9  setup_twitter_oauth(
10   consumer_key = "*****",
11   consumer_secret =
12     ↪ "*****",
13   access_token = "*****",
14   access_secret = "*****"
15 )
16 # rtweet
   ↪ =====
17 tk <- rtweet::create_token(
18   app = "SWA",
19   consumer_key = "*****",
20   consumer_secret =
21     ↪ "*****",
22   access_token =
23     ↪ "*****",
24   access_secret = "*****",
25   set_renv = FALSE

```

Listing 2: Import the twitter tokens (redacted)

```

1  n <- 1000
2  tweets.company <- search_tweets(q = 'ubisoft', n = n, token = tk,
3                                include_rts = FALSE)
4  save(tweets.company[,], file = "resources/Download_1.Rdata")

```

Listing 3: Save the Tweets to the HDD as an rdata file

```

1 (users <- unique(tweets.company$name)) %>% length()
2 x <- tweets.company$followers_count[duplicated(tweets.company$name)]
3 y <- tweets.company$friends_count[duplicated(tweets.company$name)]
4
5 ## > [1] 817

```

Listing 4: Return follower count of twitter posts

```

1 x<- rnorm(090)
2 y<- rnorm(090)
3 (xbar <- mean(x))
4 (ybar <- mean(y))
5
6 ## > [1] 4295.195
7 ## > [1] 435.9449

```

Listing 5: Determine the average number of friends and followers

```

1 (px_hat <- mean(x>xbar))
2 (py_hat <- mean(y>ybar))
3
4 ## > [1] 0.0244798
5 ## > [1] 0.2729498

```

Listing 6: Calculate the proportion of users with above average follower counts

This shows that the population follower counts is a non-normal skew-right distribution, which is expected because the number of friends is an integer value bound by zero [nist2013].

```
1  ## Resample the Data
2  kt_pop <- sample(x, size = 10^6, replace = TRUE)
3
4  ## Make the Population
5  bt_pop_data <- tibble("Followers" = bt_pop)
6  ggplot(data = bt_pop_data, aes(x = Followers)) +
7    geom_histogram(aes(y = ..density..), fill = "lightblue", bins = 35,
8      ↪ col = "pink") +
9    geom_density(col = "violetred2") +
10   scale_x_continuous(limits = c(1, 800)) +
11   theme_bw() +
12   labs(x = "Number of Followers", y = "Density",
13     title = "Bootstrapped population of Follower Numbers")
```

Listing 7: Bootstrapping a population from the sample.

References

references

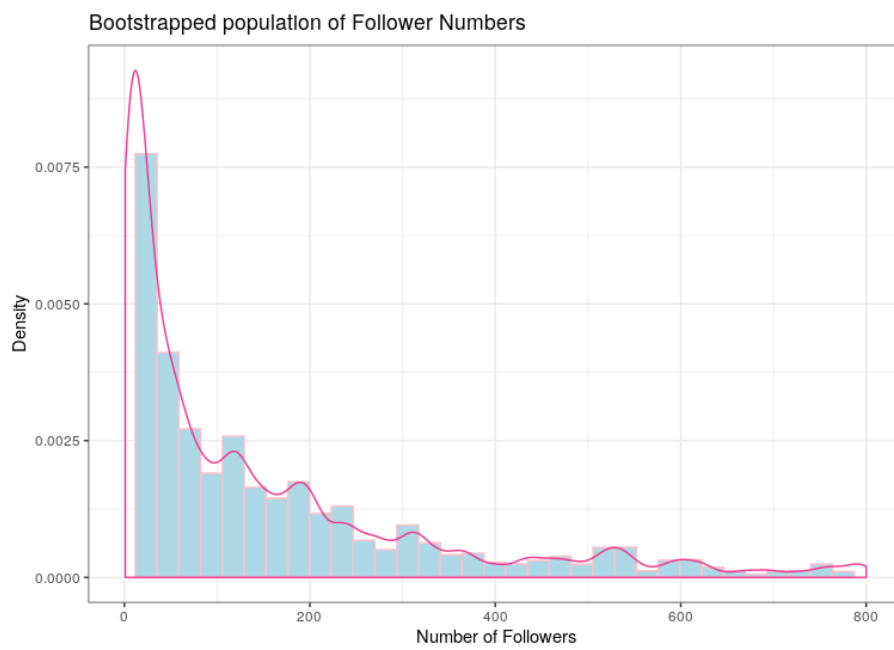


Figure 1: Histogram of the bootstrapped population of follower counts