

# Starbucks and Dunkin' Donuts Analysis

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4/6/2020

## Load in and Clean Data

```
starbucks <- read.csv("directory.csv")

# change factor variables to characters
starbucks$City <- as.character(starbucks$City)
starbucks$State.Province <- as.character(starbucks$State.Province)
starbucks$Country <- as.character(starbucks$Country)
starbucks$Postcode <- as.character(starbucks$Postcode)
starbucks$Brand <- as.character(starbucks$Brand)

# filter starbucks to only US locations, filter out alaska and hawaii
# starbucks <- starbucks[(starbucks$Country == 'US') & (starbucks$State.Province != 'A
K') & starbucks$State.Province != 'HI',]
starbucks <- starbucks %>%
  filter(Country == 'US')

# change state abbreviations to lowercase and names to match map data
starbucks$State.Province <- tolower(abbr2state(starbucks$State.Province))

# create state abbreviation column for map
starbucks$State.Abbbr <- state2abbr(starbucks$State.Province)

dunkin <- read.csv("dunkin.csv")
```

## Stock Price Analysis

```
# extract stock price data over time period (August 2011- January 2020)
#sbux <- getSymbols("SBUX", src = "yahoo", from = "2011-08-01", to = "2020-04-01", auto.
assign = FALSE)
#dnkn <- getSymbols("DNKN", src = "yahoo", from = "2011-08-01", to = "2020-04-01", auto.
assign = FALSE)

# timeout issues, just load in variables used for coffee_app (includes stock price data
scraping using the above code)
load(file = "coffee_vars.rda")
```

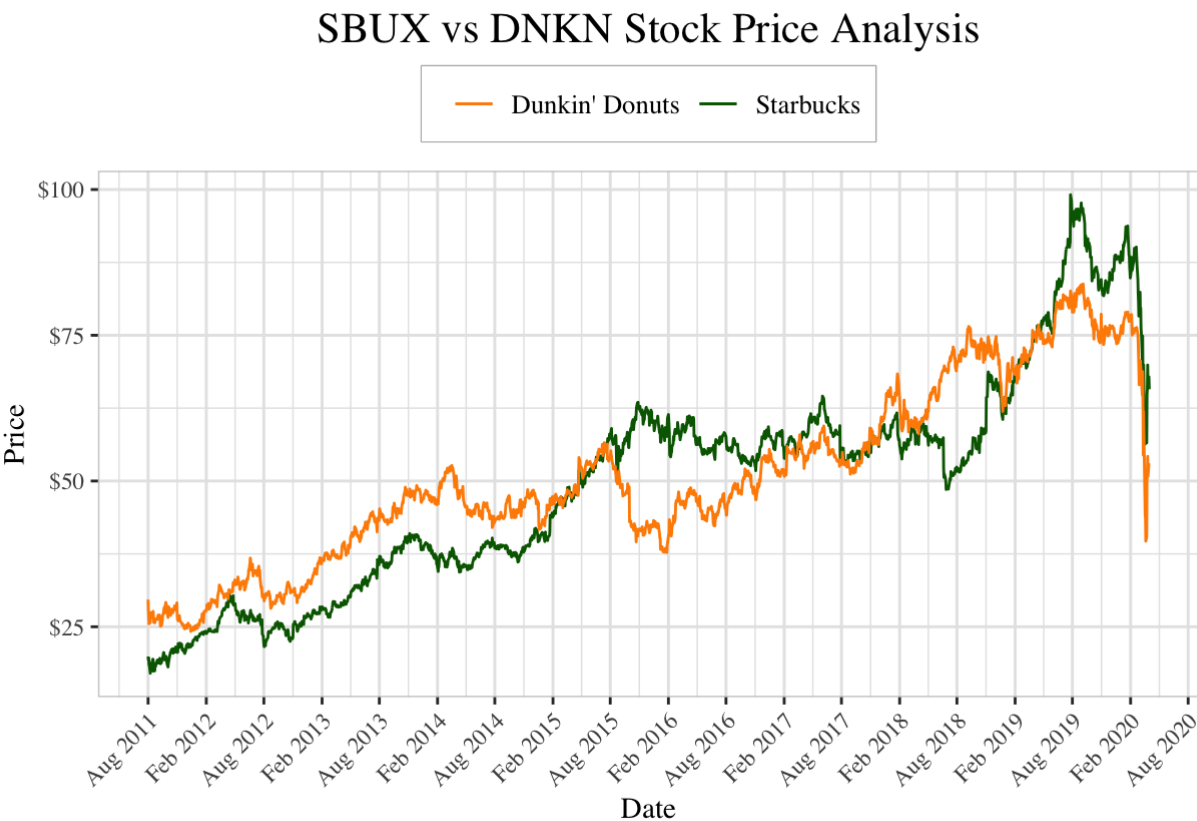
```
sbux <- as.data.frame(sbox)
dnkn <- as.data.frame(dnkn)

# create date column and column with only year for shiny app
sbux$date <- as.Date(format(rownames(sbox)))
sbux$year <- as.numeric(format(sbox$date, "%Y"))

dnkn$date <- as.Date(format(rownames(dnkn)))
dnkn$year <- as.numeric(format(dnkn$date, "%Y"))
```

```
# create plots showing the price changes
stocks_plot <- ggplot() +
  geom_line(sbox, mapping = aes(x = date, y = SBUX.Close, color = "Starbucks")) +
  geom_line(dnkn, mapping = aes(x = date, y = DNKN.Close, color = "Dunkin' Donuts")) +
  scale_color_manual(name = "", values = c("Starbucks" = "darkgreen", "Dunkin' Donuts" =
"darkorange")) +
  labs(
    x = "Date",
    y = "Price",
    title = "SBUX vs DNKN Stock Price Analysis") +
  theme_bw() +
  theme(
    text = element_text(family = "Times"),
    axis.text.x = element_text(angle = 45, hjust = 1),
    plot.title = element_text(hjust = 0.5, size = 16),
    plot.margin = unit(c(1,1,1,1), "cm"),
    legend.position = "top",
    legend.text = element_text(size = 10),
    legend.box.background = element_rect(color = "grey70"),
    panel.border = element_rect(linetype = 'solid', fill = NA, color = "grey80"),
    panel.grid = element_line(color = "grey90", linetype = "solid", lineend = "butt")) +
  scale_x_date(date_labels = "%b %Y", date_breaks = "6 months") +
  scale_y_continuous(labels=scales::dollar_format())

stocks_plot
```



# Map

```
# combine dataframes for mapping with plotly

# start by masking starbucks and dunkin dataframes, to have only the columns needed for
# mapping
# and rename the columns so they match
starbucks_map <- starbucks %>%
  select(Brand, City, State.Abbbr, Country, Longitude, Latitude)

dunkin_map <- dunkin %>%
  select(name, city, state, country, lng, lat) %>%
  rename(Brand = name, City = city, State.Abbbr = state, Country = country, Longitude = lng, Latitude = lat)

# rbind them into a new dataframe
map_data <- rbind(starbucks_map, dunkin_map)

# filter for only starbucks and dunkin donuts brands (remove Teavana, Coffee House Holdings, and Evolution Fresh)
# and add color column for mapping
# and add city.state column for hovertext
map_data <- map_data %>%
  filter(Brand %in% c("Starbucks", "Dunkin Donuts")) %>%
  mutate(Color = ifelse(Brand == "Starbucks", "darkgreen",
                        ifelse(Brand == "Dunkin Donuts", "orange", NA))) %>%
  mutate(City.State = paste(City, State.Abbbr, sep = ', '))

head(map_data)
```

```
##      Brand      City State.Abbbr Country Longitude Latitude      Color
## 1 Starbucks Anchorage      AK      US    -149.78     61.21 darkgreen
## 2 Starbucks Anchorage      AK      US    -149.84     61.14 darkgreen
## 3 Starbucks Anchorage      AK      US    -149.85     61.11 darkgreen
## 4 Starbucks Anchorage      AK      US    -149.89     61.13 darkgreen
## 5 Starbucks Anchorage      AK      US    -149.86     61.14 darkgreen
## 6 Starbucks Anchorage      AK      US    -149.87     61.19 darkgreen
##      City.State
## 1 Anchorage, AK
## 2 Anchorage, AK
## 3 Anchorage, AK
## 4 Anchorage, AK
## 5 Anchorage, AK
## 6 Anchorage, AK
```

```

#plotly map with streetmap style

# list with colors orange and dark green to assign below
map_colors <- c("#FF8000", "#006600")

spatial <- map_data
spatial <- spatial %>%
  plot_ly(
    lat = ~Latitude,
    lon = ~Longitude,
    color = ~Brand, # color by Brand
    colors = map_colors, # assign map_colors list as colors to use
    marker = list(opacity = .4), # set opacity (alpha) to .4 to help show density better
    type = 'scattermapbox',
    hovertext = map_data[, "City.State"]) # use new column with City and State for hover
ext

# assign layout and style
spatial <- spatial %>%
  layout(
    mapbox = list(
      style = 'open-street-map',
      zoom = 2.5,
      center = list(lon = -98, lat = 34)))

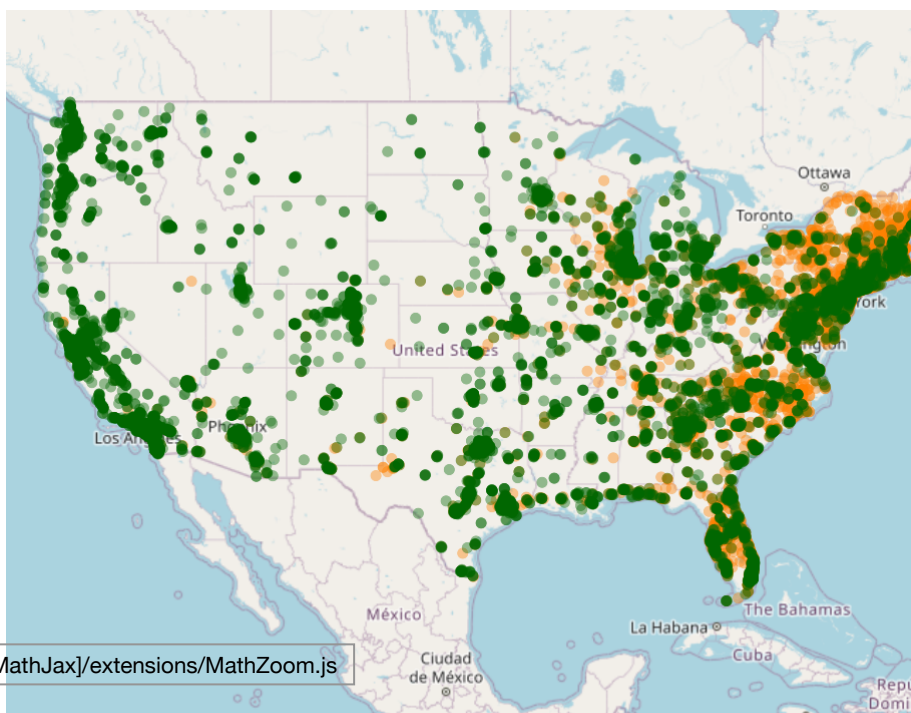
spatial

```

```

## No scattermapbox mode specified:
##   Setting the mode to markers
##   Read more about this attribute -> https://plot.ly/r/reference/#scatter-mode

```



● Dunkin Donuts  
 ● Starbucks

Loading [MathJax]/extensions/MathZoom.js



## Summary Table

```
# create summary that gets the number of stores for each brand in each state
summary <- map_data %>%
  select(Brand, State.Abbbr) %>%
  group_by(Brand, State.Abbbr) %>%
  summarize(
    stores = length(State.Abbbr)
  ) %>%
  rename(State = State.Abbbr)

# use dcast to "flip" the summary and have the states and brands as columns, while store
s is the values
# use adorn_totals to create a row at the bottom with total number of stores in the USA
for each brand
summary_table <- dcast(summary, State~Brand, value.var = "stores")
summary_table <- summary_table %>%
  adorn_totals("row")

summary_table
```

##	State	Dunkin Donuts	Starbucks
##	AK	2	48
##	AL	41	84
##	AR	8	54
##	AZ	96	479
##	CA	116	2782
##	CO	45	477
##	CT	524	118
##	DC	26	91
##	DE	67	24
##	FL	1002	671
##	GA	228	318
##	HI	12	98
##	IA	33	87
##	ID	NA	66
##	IL	695	562
##	IN	87	215
##	KS	29	93
##	KY	38	114
##	LA	14	82
##	MA	1156	262
##	MD	282	252
##	ME	161	29
##	MI	85	272
##	MN	23	175
##	MO	48	181
##	MS	8	32
##	MT	NA	36
##	NC	344	330
##	ND	NA	13
##	NE	19	56
##	NH	220	26
##	NJ	892	249
##	NM	16	75
##	NV	26	249
##	NY	1501	627
##	OH	211	366
##	OK	11	77
##	OR	NA	356
##	PA	647	348
##	RI	175	26
##	SC	119	129
##	SD	NA	24
##	TN	106	175
##	TX	175	1024
##	UT	7	97
##	VA	263	421
##	VT	52	7
##	WA	NA	747
##	WI	64	140
##	WV	23	24
##	WY	1	23
##	DC	26	91
##	PR	NA	NA
##	VI	NA	NA
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##	VI	NA	NA
##	GU	NA	NA
##	AS	NA	NA
##	FM	NA	NA
##	GU	NA	NA
##	HI	12	98
##	IA	33	87
##	ID	NA	66

# Google Trends Analysis

Hits is a range from 0 to 100 in relation to the search term's most popular time in the range provided. So, a hits level of 40 means the term is being searched at 40% of the highest its ever been searched.

```
# starbucks gtrendR scraping
sbux_food1 <- c("starbucks breakfast", "starbucks lunch", "starbucks sandwich", "starbucks muffin")

sbux_food2 <- c("starbucks bagel", "starbucks bakery", "starbucks beyond meat", "starbucks snacks",
               "starbucks food")

sbux_drinks1 <- c("starbucks latte", "starbucks americano", "starbucks cappuccino", "starbucks macchiato",
                 "starbucks espresso")

sbux_drinks2 <- c("starbucks cold brew", "starbucks iced coffee", "starbucks iced latte",
                 "starbucks coffee",
                 "starbucks drinks")

sbux_drinks3 <- c("starbucks hot chocolate", "starbucks tea", "starbucks chai", "starbucks dark roast",
                 "starbucks medium roast")

# run gtrends, use same time frame as stock analysis
# sbux_food_trends1 <- gtrends(keyword = sbux_food1, gprop = "web", geo = "US", time = "2011-08-01 2020-04-01")
# sbux_food_trends2 <- gtrends(keyword = sbux_food2, gprop = "web", geo = "US", time = "2011-08-01 2020-04-01")
# sbux_drinks_trends1 <- gtrends(keyword = sbux_drinks1, gprop = "web", geo = "US", time = "2011-08-01 2020-04-01")
# sbux_drinks_trends2 <- gtrends(keyword = sbux_drinks2, gprop = "web", geo = "US", time = "2011-08-01 2020-04-01")
# sbux_drinks_trends3 <- gtrends(keyword = sbux_drinks3, gprop = "web", geo = "US", time = "2011-08-01 2020-04-01")

# timeout issues running this as well, variables are stored in loaded rda file "coffee_v ars", data was scraped using the code above
```



```
# dunkin gtrendR scraping
dnkn_food1 <- c("dunkin breakfast", "dunkin lunch", "dunkin sandwich", "dunkin muffin")

dnkn_food2 <- c("dunkin bagel", "dunkin bakery", "dunkin beyond meat", "dunkin snacks",
               "dunkin food")

dnkn_drinks1 <- c("dunkin latte", "dunkin americano", "dunkin cappucino", "dunkin macchiato",
                 "dunkin espresso")

dnkn_drinks2 <- c("dunkin cold brew", "dunkin iced coffee", "dunkin iced latte", "dunkin coffee",
                 "dunkin drinks")

dnkn_drinks3 <- c("dunkin hot chocolate", "dunkin tea", "dunkin chai", "dunkin dark roast",
                 "dunkin medium roast")

# run gtrends, use same time frame as stock analysis
# dnkn_food_trends1 <- gtrends(keyword = dnkn_food1, gprop = "web", geo = "US", time =
#   "2011-08-01 2020-04-01")
# dnkn_food_trends2 <- gtrends(keyword = dnkn_food2, gprop = "web", geo = "US", time =
#   "2011-08-01 2020-04-01")
# dnkn_drinks_trends1 <- gtrends(keyword = dnkn_drinks1, gprop = "web", geo = "US", time =
#   "2011-08-01 2020-04-01")
# dnkn_drinks_trends2 <- gtrends(keyword = dnkn_drinks2, gprop = "web", geo = "US", time =
#   "2011-08-01 2020-04-01")
# dnkn_drinks_trends3 <- gtrends(keyword = dnkn_drinks3, gprop = "web", geo = "US", time =
#   "2011-08-01 2020-04-01")

# same as starbucks google data, timeout issues running this as well, variables are stored in loaded rda file "coffee_vars", data was scraped using the code above
```

```

# merge starbucks and dunkin into one dataframe each
# select interest_over_time gtrendsR result
# create column with month number, only plot every six months or twice a year

## THIS DATA CLEANING CAN BE SEEN IN FINAL GOOGLE TRENDS DATAFRAMES
## IT IS COMMENTED OUT DUE TO TIMEOUT ISSUES IN ABOVE CHUNKS, BUT THE CHANGES WERE SAVED
AND LOADED IN WITH "coffee_vars.rda"
# sbux_food_trends <- rbind(sbxux_food_trends1$interest_over_time, sbux_food_trends2$inte
rest_over_time)
# sbux_drinks_trends <- rbind(sbxux_drinks_trends1$interest_over_time, sbux_drinks_trends
2$interest_over_time,
#
#                               sbux_drinks_trends3$interest_over_time)
#
# dnkn_food_trends <- rbind(dnkn_food_trends1$interest_over_time, dnkn_food_trends2$inte
rest_over_time)
# dnkn_drinks_trends <- rbind(dnkn_drinks_trends1$interest_over_time, dnkn_drinks_trends
2$interest_over_time,
#
#                               dnkn_drinks_trends3$interest_over_time)
#
# # add month column to dataframes
# sbux_food_trends <- sbux_food_trends %>%
#   mutate(
#     month = as.numeric(format(date, "%m"))
#   )
#
# sbux_drinks_trends <- sbux_drinks_trends %>%
#   mutate(
#     month = as.numeric(format(date, "%m"))
#   )
#
# dnkn_food_trends <- dnkn_food_trends %>%
#   mutate(
#     month = as.numeric(format(date, "%m"))
#   )
#
# dnkn_drinks_trends <- dnkn_drinks_trends %>%
#   mutate(
#     month = as.numeric(format(date, "%m"))
#   )
#
# # change "hits" to numeric rather than factor
# sbux_drinks_trends$hits <- as.numeric(sbxux_drinks_trends$hits)
# sbux_food_trends$hits <- as.numeric(sbxux_food_trends$hits)
# dnkn_drinks_trends$hits <- as.numeric(dnkn_drinks_trends$hits)
# dnkn_food_trends$hits <- as.numeric(dnkn_food_trends$hits)
#
#
# # set NAs to zero for hits column
# sbux_drinks_trends$hits[is.na(sbxux_drinks_trends$hits)] <- 0
# sbux_food_trends$hits[is.na(sbxux_food_trends$hits)] <- 0
# dnkn_drinks_trends$hits[is.na(dnkn_drinks_trends$hits)] <- 0
# dnkn_food_trends$hits[is.na(dnkn_food_trends$hits)] <- 0

```

```
#
#
# # create original column to preserve original keyword column
# # use separate split keyword column into brand and item
# # use unite to concatenate items with two words back together
#
# sbux_drinks_trends <- sbux_drinks_trends %>%
#   mutate(
#     original = keyword
#   ) %>%
#   separate(keyword, c("brand", "item1", "item2"), sep = " ") %>%
#   unite("item_total", item1, item2, sep = " ", na.rm = TRUE)
#
# sbux_food_trends <- sbux_food_trends %>%
#   mutate(
#     original = keyword
#   ) %>%
#   separate(keyword, c("brand", "item1", "item2"), sep = " ") %>%
#   unite("item_total", item1, item2, sep = " ", na.rm = TRUE)
#
# dnkn_drinks_trends <- dnkn_drinks_trends %>%
#   mutate(
#     original = keyword
#   ) %>%
#   separate(keyword, c("brand", "item1", "item2"), sep = " ") %>%
#   unite("item_total", item1, item2, sep = " ", na.rm = TRUE)
#
# dnkn_food_trends <- dnkn_food_trends %>%
#   mutate(
#     original = keyword
#   ) %>%
#   separate(keyword, c("brand", "item1", "item2"), sep = " ") %>%
#   unite("item_total", item1, item2, sep = " ", na.rm = TRUE)
#
# # use stringr to make all item_total columns capital first letters
#
# sbux_drinks_trends$item_total <- str_to_title(sbx_drinks_trends$item_total)
# sbux_food_trends$item_total <- str_to_title(sbx_food_trends$item_total)
# dnkn_drinks_trends$item_total <- str_to_title(dnkn_drinks_trends$item_total)
# dnkn_food_trends$item_total <- str_to_title(dnkn_food_trends$item_total)
#
```

```

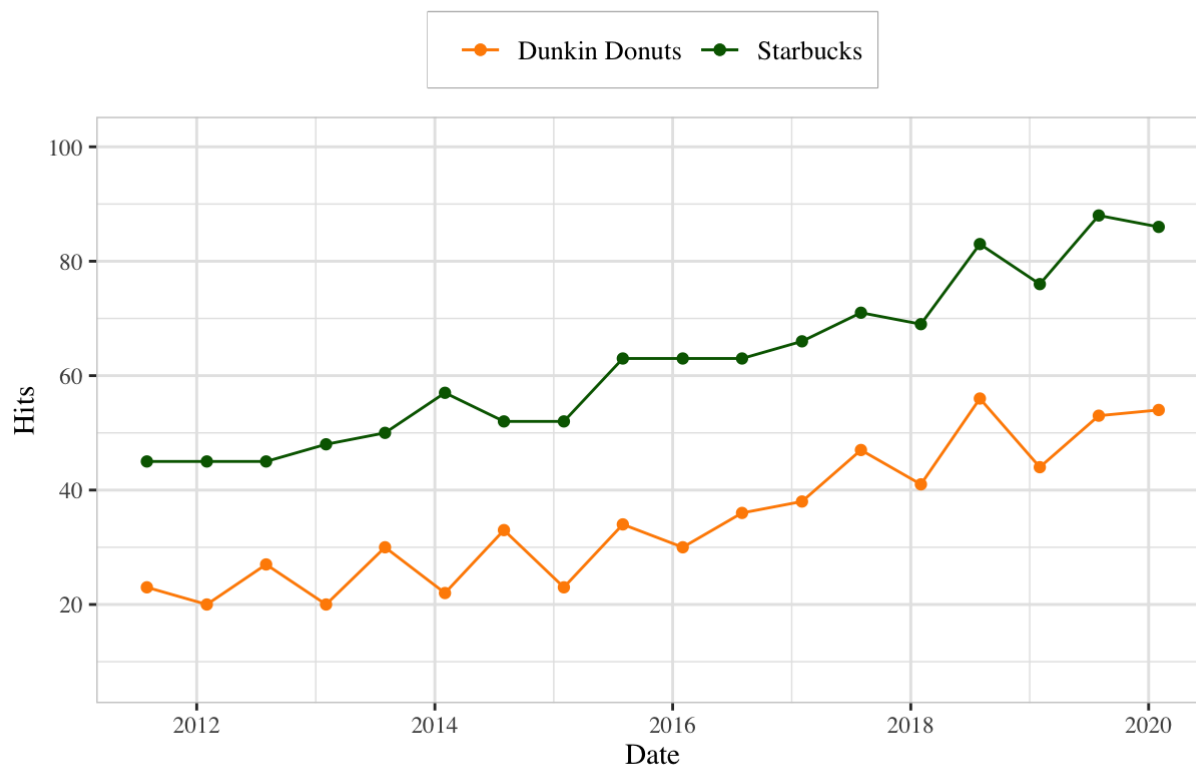
# example of plot
sbux_coffee <- sbux_drinks_trends %>%
  filter(item_total == "Coffee" & month %in% c(2,8))

dnkn_coffee <- dnkn_drinks_trends %>%
  filter(item_total == "Coffee" & month %in% c(2,8))

ggplot() +
  geom_point(sbux_coffee, mapping = aes(x = date, y = hits, color = "Starbucks")) +
  geom_line(sbux_coffee, mapping = aes(x = date, y = hits, color = "Starbucks", group =
1)) +
  geom_point(dnkn_coffee, mapping = aes(x = date, y = hits, color = "Dunkin Donuts")) +
  geom_line(dnkn_coffee, mapping = aes(x = date, y = hits, color = "Dunkin Donuts", grou
p = 1)) +
  scale_color_manual(name = "", values = c("Starbucks" = "darkgreen", "Dunkin Donuts" =
"darkorange")) +
  labs(
    x = "Date",
    y = "Hits",
    title = "Coffee"
  ) +
  theme_bw() +
  theme(
    text = element_text(family = "Times"),
    plot.title = element_text(hjust = 0.5, size = 16),
    plot.margin = unit(c(1,1,1,1), "cm"),
    legend.position = "top",
    legend.text = element_text(size = 10),
    legend.box.background = element_rect(color = "grey70"),
    panel.border = element_rect(linetype = 'solid', fill = NA, color = "grey80"),
    panel.grid = element_line(color = "grey90", linetype = "solid", lineend = "butt"),
  ) +
  scale_y_continuous(breaks = seq(0,100, by = 20), expand = c(.25,.25))

```

## Coffee



**Save dataframes and other variables needed for shiny app as an rda file that can be loaded in**

```
# save(sbox, dnkn, map_data, summary_table, sbux_drinks_trends, sbux_food_trends, dnkn_drinks_trends,
#       dnkn_food_trends, file = "coffee_vars.rda" )
```

# References

*For Starbucks and Dunkin Donuts Data*

<https://www.kaggle.com/starbucks/store-locations> (<https://www.kaggle.com/starbucks/store-locations>)

<https://www.kaggle.com/jpbulman/usa-dunkin-donuts-stores> (<https://www.kaggle.com/jpbulman/usa-dunkin-donuts-stores>)

*For Flat json File Conversion to csv Table*

<https://konklone.io/json/> (<https://konklone.io/json/>)

*For Stock Price Info (scraped with quantmod package)*

<https://finance.yahoo.com/> (<https://finance.yahoo.com/>)

*For Google Trends Data (scraped with gtrendsR package)*

<https://trends.google.com/trends/?geo=US> (<https://trends.google.com/trends/?geo=US>)