

MATH 158 - Data Description and Descriptive Statistics

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Introduction to Data

The data from this project comes from TidyTuesday and fivethirtyeight on GitHub. This data is a collection of the top 10 companies who posted the most advertisements during the Superbowl between the years 2000 and 2020. Because of errors in data collection, seventeen of the videos had to be removed from the data. From this, we want to compare the different qualities in the video to the youtube performance after the event.

The observational unit is each individual advertisement, and there are 10 variables. The categorical variables are logical binary variables which identify as True or False.

Quantitative Data

- Year: This variable indicates the year the Superbowl advertisement aired on TV.
- View Count: This variable indicates how many youtube views the advertisement has received.
- Like Count: This variable indicates how many youtube likes the advertisement has received.
- Dislike Count: This variable indicates how many youtube dislikes the advertisement has received.
- Comment Count: This variable indicates how many youtube comments the advertisement has received.

Categorical Data

- Funny: This variable indicates whether the advertisement is intended to be funny.
- Celebrity: This variable indicates whether a celebrity is in the advertisement.
- Danger: This variable indicates whether there is danger in the advertisement.
- Animals: This variable indicates whether there are animals in the advertisement.
- Use Sex: This variable indicates whether there is use of sexuality in the advertisement.

Summary of Statistics

As mentioned above, the dataset looked at the 10 companies who prepared advertisements between 2000 and 2020. Out of these 10 companies, this is how many total advertisements they posted each year.

```
maindata <- read.csv("Data - ExportedData.csv")
my_data <- as.data.frame(maindata)
ggplot(my_data, aes(x=year)) + geom_bar()
```

This next graph will show how many advertisements each company created for the span of Superbowls.

```
maindata <- read.csv("Data - ExportedData.csv")
my_data <- as.data.frame(maindata)
ggplot(my_data, aes(x=brand)) + geom_bar()
```

For this next graph, I indicated some critical statistics encompassing all of the videos for each brand.

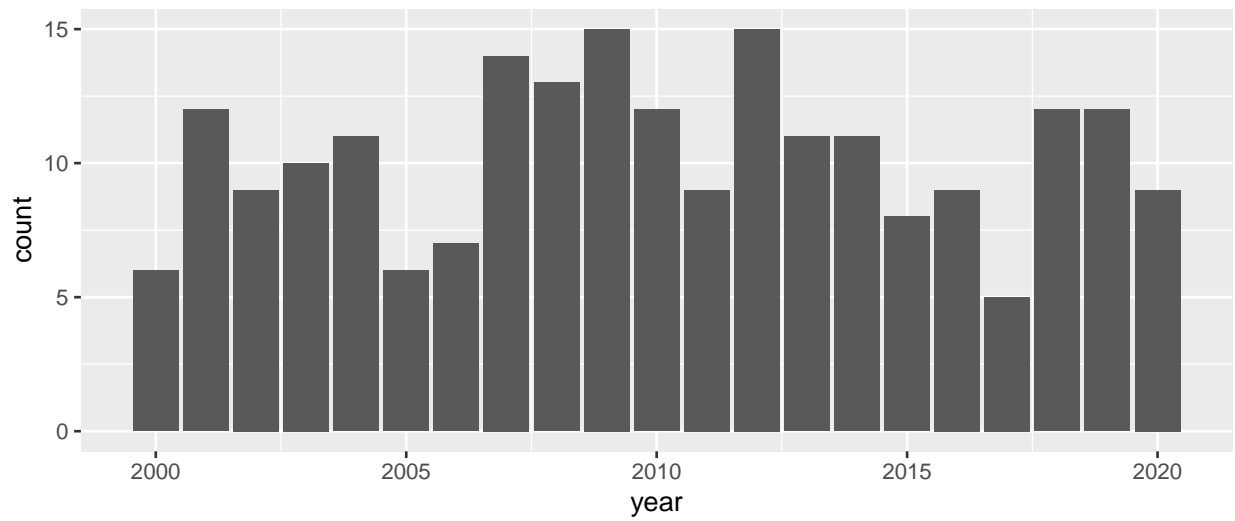


Figure 1: Advertisements by Year

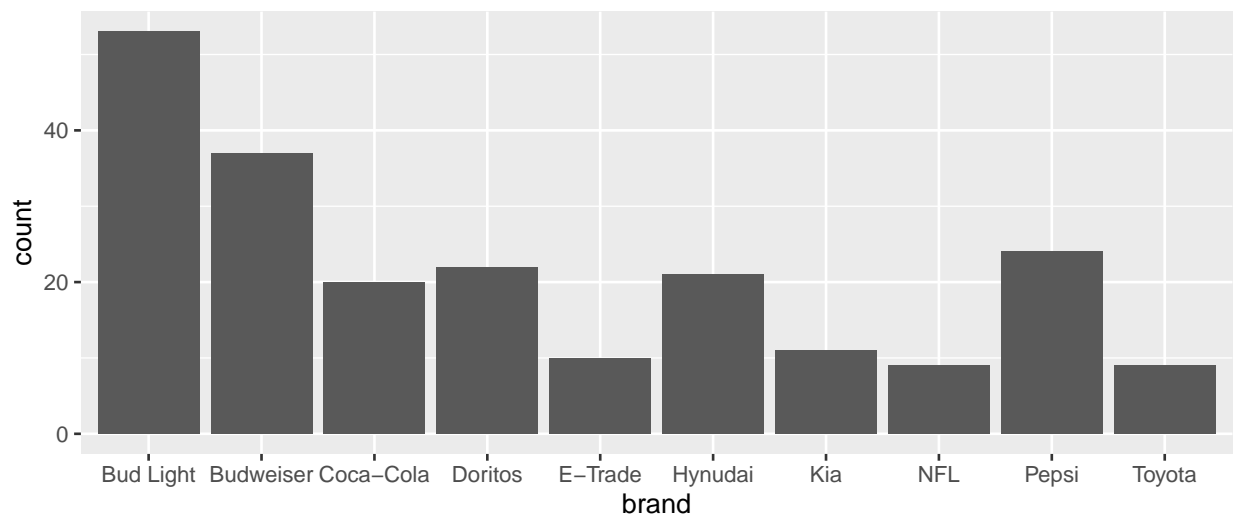


Figure 2: Advertisements by Brand

```

maindata <- read.csv("Data - ExportedData.csv")
my_data <- as.data.frame(maindata)
df6<-maindata %>%
  group_by(brand) %>%
  summarise(Min = min(view_count),
            Max=max(view_count),
            mean = round(mean(view_count)),
            median = round(median(view_count)))
df6

```

```

## # A tibble: 10 x 5
##   brand      Min      Max    mean median
##   <fct>    <int>    <int>  <dbl> <dbl>
## 1 Bud Light   125  7658201 262701 34565
## 2 Budweiser   10  28785122 1026244 50088
## 3 Coca-Cola  179  22849816 1618888 72245
## 4 Doritos   2985 176373378 8875610 225794
## 5 E-Trade     21  1046640 172402 50811
## 6 Hynudai     56   373684  44565  5049
## 7 Kia        518    87687   30657 17892
## 8 NFL       18670 26727063 4097798 403641
## 9 Pepsi      111   669906  121456 49830
## 10 Toyota   4873   353513  106807 32091

```

For this table, I want to reveal the number of observations which identify as True or False for every categorical variable.

Object 4:

	Funny	Celebrity	Danger	Animals	Use_Sex
True	171	71	75	92	66
False	45	145	141	124	150

Next, I highlighted the mean and median number of views for advertisements based on responding True or False to each category.

Object 5:

```

df7<-maindata %>% group_by(funny) %>% summarise(mean = round(mean(view_count)),
median = round(median(view_count)))
df8<-maindata %>% group_by(celebrity) %>% summarise(mean = round(mean(view_count)),
median = round(median(view_count)))
df9<-maindata %>% group_by(danger) %>% summarise(mean = round(mean(view_count)),
median = round(median(view_count)))
df10<-maindata %>% group_by(animals) %>% summarise(mean = round(mean(view_count)),
median = round(median(view_count)))
df11<-maindata %>% group_by(use_sex) %>% summarise(mean = round(mean(view_count)),
median = round(median(view_count)))
df_1 <- merge(df7, df8, by = "row.names", all = TRUE)
df_2 <- merge(df_1, df9, by = "row.names", all = TRUE)
df_3 <- merge(df_2, df10, by = "row.names", all = TRUE)
df_4 <- merge(df_3, df11, by = "row.names", all = TRUE)
df_5 <- df_4[ -c(1:4) ]
df_5

```

```

##   funny mean.x median.x celebrity mean.y median.y danger mean.x.1 median.x.1
## 1 FALSE 1359659   40358     FALSE 1768268   48978  FALSE 1680690   39814

```

##	2	TRUE	1559676	48546	TRUE	822187	41323	TRUE	1096275	61656
##		animals	mean.y.1	median.y.1	use_sex	mean	median			
##	1	FALSE	1941964	41379	FALSE	1938213	48035			
##	2	TRUE	692933	50850	TRUE	204302	36832			