# Alcohol Abuse and Mental/Physical Health in the United States

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## Loading packages

```
library(haven)
library(sf)
library(ggplot2)
library(sjPlot)
library(sjmisc)
library(sjlabelled)
library(plyr)
library(tidyverse)
library(formatR)
```

# Importing data from the CDC's BRFSS - 2018

```
zip_tf <- tempfile()
zip_url <- "https://www.cdc.gov/brfss/annual_data/2018/files/LLCP2018XPT.zip"
download.file(zip_url, zip_tf, mode = "wb")
brfss_tbl <- read_xpt(zip_tf)
brfss_df <- data.frame(brfss_tbl)
names(brfss_df) <- tolower(names(brfss_df))
brfss_df[, "one"] <- 1</pre>
```

# Data Manipulation and Cleaning

Selecting only the variable names that we want to keep to make the data more manageable

```
variables <- c("x_state", "maxdrnks", "alcday5", "avedrnk2",
        "drnk3ge5", "menthlth", "poorhlth", "addepev2", "genhlth",
        "physhlth", "x_ageg5yr", "sex1", "one", "x_llcpwt", "x_rfbing5")
brfss_df <- brfss_df[variables]</pre>
```

#### Adding state abbreviation column

#### Subsetting the data

#### Binge Drinking Prevalence

```
binge_prev <- brfss_df[(brfss_df$x_rfbing5 == 1 | brfss_df$x_rfbing5 ==
2) & !is.na(brfss_df$x_rfbing5), ]</pre>
```

#### Binge Drinking Intensity

#### Binge Drinking Frequency

#### **Average Alcohol Consumption**

#### Average Drinking Frequency

```
# Dataframe excluding NA and 'not sure' or 'refused'
# responses from the 'number of days per week/month where
# you consumed an alcoholic beverage' column
alc_days <- brfss_df[(brfss_df$alcday5 %in% 101:107 | brfss_df$alcday5 %in%
    201:230 | brfss df$alcday5 == 888) & !is.na(brfss df$alcday5),
   1
# Replacing '888' values with '0', since a response of
# '888' means 'No drinks in the past 30 days'
alc_days$alcday5[alc_days$alcday5 == "888"] <- 0</pre>
# Remove the '2' from 201-230 values, since 2 = ays
# per month and dividing it by 7 to get 'days per week'
alc_days$alcday5 <- ifelse(alc_days$alcday5 %in% 201:209, gsub("20",
    "", alc_days$alcday5), alc_days$alcday5)
alc days$alcday5 <- ifelse(alc days$alcday5 %in% 210:230, gsub("2",
    "", alc_days$alcday5), alc_days$alcday5)
alc_days$alcday5 <- ifelse(alc_days$alcday5 %in% 1:30, round((as.numeric(alc_days$alcday5))/7,
    2), alc days$alcday5)
# Remove the '10' from 101-107 values, since 1 _ _ = days
# per week
alc_days$alcday5 <- ifelse(alc_days$alcday5 %in% 101:107, gsub("10",
    "", alc_days$alcday5), alc_days$alcday5)
```

# **Exploratory Data Analysis and Visualizations**

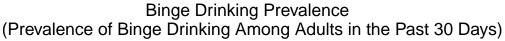
#### Binge Drinking Prevalence

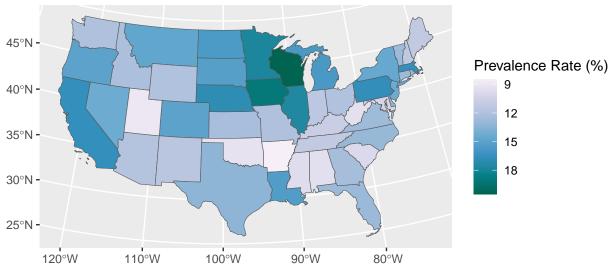
Calculating the proportion of binge drinkers in each state

```
yesNo <- data.frame(Yes = rowSums(binge_prev["x_rfbing5"] ==
2), No = rowSums(binge_prev["x_rfbing5"] == 1))
binge <- aggregate(yesNo, binge_prev["stname"], sum)
binge$percent <- (binge$Yes)/(binge$Yes + binge$No) * 100
binge$percent <- round(binge$percent, 2)</pre>
```

#### Creating a map to display the proportion of binge drinkers in each state

```
states <- st_read("2015-2019-acs-states.geojson")</pre>
## Reading layer '2015-2019-acs-states' from data source
     'C:\Users\jane9\OneDrive\Documents\March Analytics Project\2015-2019-acs-states.geojson'
     using driver 'GeoJSON'
## Simple feature collection with 52 features and 93 fields
## Geometry type: MULTIPOLYGON
## Dimension:
                  XY
## Bounding box: xmin: -179 ymin: 17.91377 xmax: -65.22157 ymax: 71.35256
## Geodetic CRS: WGS 84
states <- states[!(states$ST %in% c("PR", "AK", "HI")), ]</pre>
binge <- binge[!(binge$stname %in% c("Puerto Rico ", "Alaska",</pre>
    "Hawaii")), ]
states <- st_transform(states, 6580)</pre>
states_prevalence <- merge(states, binge[, c("stname", "percent")],</pre>
    by.x = "Name", by.y = "stname")
map <- ggplot() + geom_sf(data = states_prevalence, aes(fill = binge$percent)) +</pre>
    scale_fill_distiller(palette = "PuBuGn", trans = "reverse") +
    labs(title = "Binge Drinking Prevalence \n (Prevalence of Binge Drinking Among Adults in the Past 3
        fill = "Prevalence Rate (%)") + theme(plot.title = element_text(hjust = 0.5))
print(map)
```



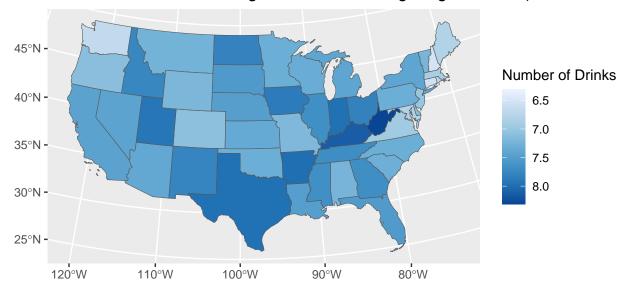


## Binge Drinking Intensity

Looking at state means and quartiles of 'most drinks consumed on a single occasion in the past 30 days' among those who engaged in binge drinking in the past 30 days

#### Creating a map to display the state means for binge drinking intensity

# Binge Drinking Intensity Max Number of Drinks on a Single Occasion Among Binge Drinkers)



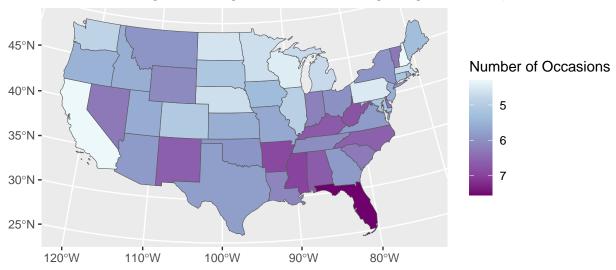
### Binge Drinking Frequency

Calculating state means for binge drinking frequency among those who engage in binge drinking

```
# Creating a subset of the data, which only includes those
# who have engaged in binge drinking in the past 30 days
```

#### Creating a map displaying binge drinking frequency

Binge Drinking Frequency
1ean Number of Binge Drinking Occasions Among Binge Drinkers)



## Depressive Disorders and Mental/Physical Health Condition vs Binge Drinking

Constructing a Logistic Regression Model to Predict Whether a Survey Respondent is a Binge Drinker Based on Their Health Condition

Preparing the Data for Logistic Regression

 $\begin{tabular}{l} \textbf{Logistic Regression Model: Binge Drinker} \sim \textbf{Depressive Disorder} + \textbf{Mental Health} + \textbf{Physical Health} + \textbf{General Health} \\ \end{tabular}$ 

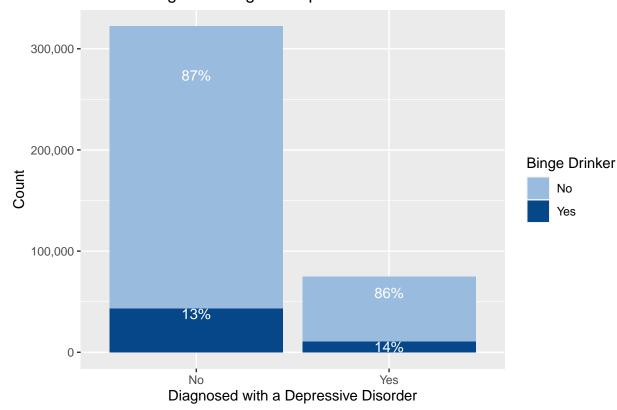
```
model <- glm(x_rfbing5 ~ addepev2 + menthlth + physhlth + genhlth,
   family = binomial(link = "logit"), data = health_prev)
summary(model)
##
## Call:
## glm(formula = x_rfbing5 ~ addepev2 + menthlth + physhlth + genhlth,
      family = binomial(link = "logit"), data = health_prev)
##
##
## Coefficients:
##
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.5511005 0.0124565 -124.52 <2e-16 ***
## addepev21 0.0213137 0.0132365
                                    1.61
                                            0.107
## menthlth
              0.0259187 0.0006286
                                   41.23
                                           <2e-16 ***
## physhlth -0.0222096 0.0007225 -30.74
                                           <2e-16 ***
             ## genhlth
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
      Null deviance: 315952 on 396886 degrees of freedom
## Residual deviance: 312393 on 396882 degrees of freedom
## AIC: 312403
## Number of Fisher Scoring iterations: 5
# Displaying results of logistic regression as a table
tab_model(model, dv.labels = "Binge Drinker (No = 0, Yes = 1)",
   pred.labels = c("(Intercept)", "Depressive Disorder", "Mental Health",
       "Physical Health", "General Health"), show.est = TRUE,
   show.stat = TRUE)
```

## Profiled confidence intervals may take longer time to compute.

## Use 'ci\_method="wald"' for faster computation of CIs.

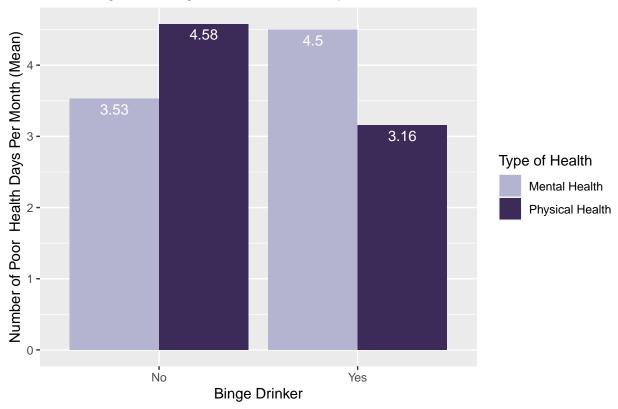
#### Stacked Bar Plot Showing Binge Drinking vs. Deppressive Disorder

## Binge Drinking vs. Depressive Disorder



Binge Drinking vs. Poor Mental and Physical Health

## Binge Drinking vs. Mental and Physical Health



#### Binge Drinking vs. General Health

```
dat3 <- aggregate(genhlth ~ x_rfbing5, health_prev, median)
ggplot(data = dat3, aes(x = x_rfbing5, y = genhlth, fill = x_rfbing5)) +
    geom_bar(stat = "identity", width = 0.5) + scale_x_discrete(labels = c("No",
    "Yes")) + scale_fill_manual(values = c("#a7d4bf", "#3a6953"),
    labels = c("No", "Yes")) + labs(title = "Binge Drinking vs. General Health",
    fill = "Binge Drinker", x = "Binge Drinker", y = "Median General Health Rating (1-5)") +</pre>
```

```
geom_text(aes(label = round(genhlth, 2)), hjust = 0.5, vjust = 4,
size = 4, color = "white") + theme(plot.title = element_text(hjust = 0.5))
```



