Final Report

due November 16, 2021 by 11:59 PM

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```
install.packages("taRifx")
install.packages("fastDummies")
library(tidyverse)
library(dplyr)
library(taRifx)
library(fastDummies)
library(infer)
library(parsnip)
drug <- readr::read_csv("Drug_Consumption.csv")
```

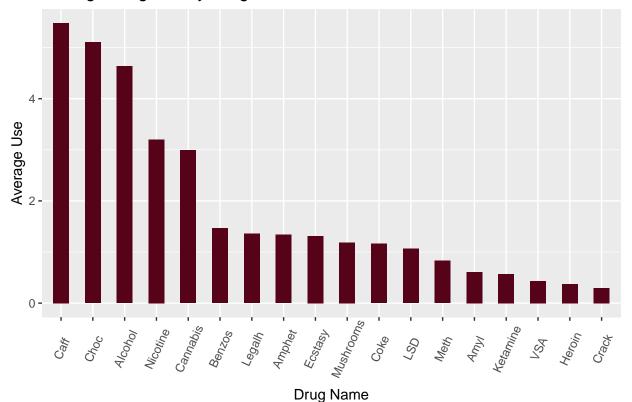
Abstract:

Background and Significance:

Methods: a) Data Collection and Variables

b) Exploratory Data Analysis

Average Drug Use by Drug



```
numdrug <- drug1 %>%
   mutate(Age = replace(Age, Age == "18-24", 0), Age = replace(Age, Age == "25-34", 1), Age = replace(Age)
numdrug <- mutate_all(numdrug, function(x) as.numeric(as.character(x)))
numdrug2 = select(numdrug, -Alcohol, -Amphet, -Amyl, -Benzos, -Caff, -Cannabis, -Choc, -Coke, -Crack, -fead(numdrug2)</pre>
```

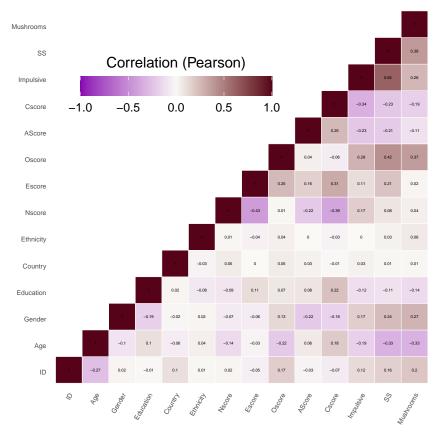
```
## # A tibble: 6 x 14
##
             Age Gender Education Country Ethnicity Nscore Escore Oscore AScore
##
     <dbl> <dbl>
                  <dbl>
                            <dbl>
                                     <dbl>
                                               <dbl> <dbl> <dbl>
                                                                     <dbl>
                                                                            <dbl>
## 1
         2
               1
                                        5
                                                   6 -0.678 1.94
                                                                    1.44
                                                                             0.761
## 2
         3
               2
                                5
                                        5
                                                   6 -0.467 0.805 -0.847 -1.62
                      1
## 3
         4
               0
                      0
                                7
                                         5
                                                   6 -0.149 -0.806 -0.0193 0.590
## 4
         5
               2
                      0
                                8
                                        5
                                                   6 0.735 -1.63 -0.452 -0.302
## 5
               5
                                3
                                        1
                                                   6 -0.678 -0.300 -1.56
         7
                                7
                                        6
                                                   6 -0.467 -1.09 -0.452
                                                                           -0.302
                      1
## # ... with 4 more variables: Cscore <dbl>, Impulsive <dbl>, SS <dbl>,
```

Mushrooms <dbl>

correlation_matrix <- round(cor(numdrug2),2)
head(correlation_matrix)</pre>

```
## ID Age Gender Education Country Ethnicity Nscore Escore Oscore
## ID 1.00 -0.27 0.02 -0.01 0.10 0.01 0.02 -0.05 0.17
## Age -0.27 1.00 -0.10 0.10 -0.06 0.04 -0.14 -0.03 -0.22
```

```
0.02 -0.10 1.00
## Gender
                                    -0.19
                                            -0.02
                                                       0.02 -0.07 -0.06
                                                                             0.13
## Education -0.01 0.10 -0.19
                                     1.00
                                             0.02
                                                       -0.08 -0.09
                                                                      0.11
                                                                             0.07
              0.10 -0.06 -0.02
                                     0.02
                                                                      0.00
## Country
                                             1.00
                                                      -0.03
                                                               0.05
                                                                             0.05
## Ethnicity 0.01 0.04
                         0.02
                                    -0.08
                                                        1.00
                                                               0.01 -0.04
                                            -0.03
                                                                             0.04
             AScore Cscore Impulsive
                                        SS Mushrooms
## ID
              -0.03 -0.07
                                0.12 0.16
                                                0.20
                     0.18
                               -0.19 - 0.33
                                               -0.33
## Age
               0.06
                                0.17 0.24
              -0.22 -0.18
## Gender
                                                0.27
                               -0.12 -0.11
## Education
               0.08
                     0.22
                                               -0.14
## Country
               0.03 -0.01
                                0.03 0.01
                                                0.01
## Ethnicity
               0.00 -0.03
                                0.00 0.03
                                                0.06
  get_upper_tri<-function(correlation_matrix){</pre>
    correlation_matrix[lower.tri(correlation_matrix)] <- NA</pre>
   return(correlation_matrix)
  }
  upper_tri <- get_upper_tri(correlation_matrix)</pre>
library(reshape2)
##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
##
melted_cormat <- melt(upper_tri, na.rm = TRUE)</pre>
library(ggplot2)
ggplot(data = melted_cormat, aes(Var2, Var1, fill = value))+
geom tile(color = "white")+
scale fill gradient2(low = "#8a02b2", high = "#560219", mid = "#FAF9F6",
  midpoint = 0, limit = c(-1,1), space = "Lab",
  name="Correlation (Pearson)") +
 theme minimal()+
theme(axis.text.x = element_text(angle = 60, vjust = 1,
    size = 5, hjust = 1), axis.text.y = element text(vjust = 1, size = 5, hjust = 1))+
 coord_fixed() +
geom_text(aes(Var2, Var1, label = value), color = "black", size = 1) +
theme(
  axis.title.x = element_blank(),
  axis.title.y = element_blank(),
  panel.grid.major = element_blank(),
  panel.border = element_blank(),
  panel.background = element_blank(),
  axis.ticks = element_blank(),
  legend.justification = c(1, 0),
  legend.position = c(0.6, 0.7),
  legend.direction = "horizontal",
  legend.key.size = unit(0.5, 'cm'))+
  guides(fill = guide_colorbar(barwidth = 10, barheight = 1,
                title.position = "top", title.hjust = 0.5))
```

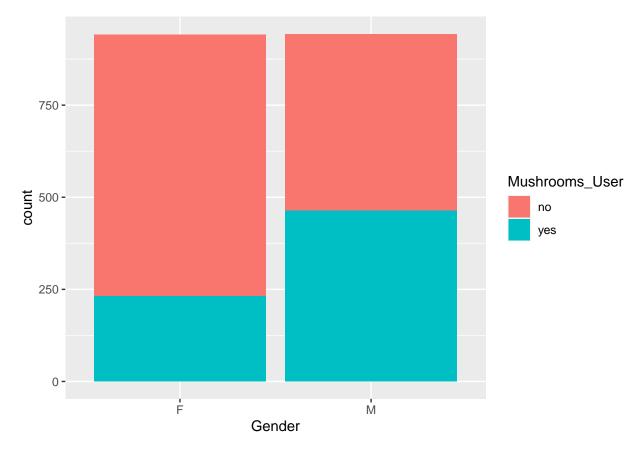


```
drug clean <- numdrug %>%
  mutate(Alcohol_User = as.factor(ifelse(Alcohol > 1, "yes", "no")),
         Amphetamine User = as.factor(ifelse(Amphet > 1, "yes", "no")),
         AmylNitrite_User = as.factor(ifelse(Amyl > 1, "yes", "no")),
         Benzos User = as.factor(ifelse(Benzos > 1, "yes", "no")),
         Caffeine User = as.factor(ifelse(Caff > 1, "yes", "no")),
         Cannabis User = as.factor(ifelse(Cannabis > 1, "yes", "no")),
         Chocolate_User = as.factor(ifelse(Choc > 1, "yes", "no")),
         Cocaine_User = as.factor(ifelse(Coke > 1, "yes", "no")),
         Crack_User = as.factor(ifelse(Crack > 1, "yes", "no")),
         Ecstasy_User = as.factor(ifelse(Ecstasy > 1, "yes", "no")),
         Heroine_User = as.factor(ifelse(Heroin > 1, "yes", "no")),
         Ketamine_User = as.factor(ifelse(Ketamine > 1, "yes", "no")),
         LegalHighs_User = as.factor(ifelse(Legalh > 1, "yes", "no")),
         LSD_User = as.factor(ifelse(LSD > 1, "yes", "no")),
         Meth_User = as.factor(ifelse(Meth > 1, "yes", "no")),
         Mushrooms_User = as.factor(ifelse(Mushrooms > 1, "yes", "no")),
         Nicotine User = as.factor(ifelse(Nicotine > 1, "yes", "no")),
         Semeron_User = as.factor(ifelse(Semer > 1, "yes", "no")),
         VSA User = as.factor(ifelse(VSA > 1, "yes", "no")))
drug_clean_2 <- drug1 %>%
   mutate(Alcohol_User = as.factor(ifelse(Alcohol > 1, "yes", "no")),
         Amphetamine User = as.factor(ifelse(Amphet > 1, "yes", "no")),
         AmylNitrite_User = as.factor(ifelse(Amyl > 1, "yes", "no")),
         Benzos_User = as.factor(ifelse(Benzos > 1, "yes", "no")),
```

```
Caffeine_User = as.factor(ifelse(Caff > 1, "yes", "no")),
         Cannabis_User = as.factor(ifelse(Cannabis > 1, "yes", "no")),
         Chocolate_User = as.factor(ifelse(Choc > 1, "yes", "no")),
         Cocaine User = as.factor(ifelse(Coke > 1, "yes", "no")),
         Crack_User = as.factor(ifelse(Crack > 1, "yes", "no")),
         Ecstasy_User = as.factor(ifelse(Ecstasy > 1, "yes", "no")),
         Heroine_User = as.factor(ifelse(Heroin > 1, "yes", "no")),
         Ketamine User = as.factor(ifelse(Ketamine > 1, "yes", "no")),
         LegalHighs User = as.factor(ifelse(Legalh > 1, "yes", "no")),
         LSD_User = as.factor(ifelse(LSD > 1, "yes", "no")),
         Meth_User = as.factor(ifelse(Meth > 1, "yes", "no")),
         Mushrooms_User = as.factor(ifelse(Mushrooms > 1, "yes", "no")),
         Nicotine_User = as.factor(ifelse(Nicotine > 1, "yes", "no")),
         Semeron_User = as.factor(ifelse(Semer > 1, "yes", "no")),
         VSA_User = as.factor(ifelse(VSA > 1, "yes", "no")))
drug byuse <- numdrug %>%
  mutate(Alcohol_User = ifelse(Alcohol > 1, 1, 0),
         Amphetamine_User = ifelse(Amphet > 1, 1, 0),
         AmylNitrite_User = ifelse(Amyl > 1, 1, 0),
         Benzos_User = ifelse(Benzos > 1, 1, 0),
         Caffeine_User = ifelse(Caff > 1, 1, 0),
         Cannabis_User = ifelse(Cannabis > 1, 1, 0),
         Chocolate_User = ifelse(Choc > 1, 1, 0),
         Cocaine_User = ifelse(Coke > 1, 1, 0),
         Crack_User = ifelse(Crack > 1, 1, 0),
         Ecstasy_User = ifelse(Ecstasy > 1, 1, 0),
         Heroine User = ifelse(Heroin > 1, 1, 0),
         Ketamine User = ifelse(Ketamine > 1, 1, 0),
         LegalHighs User = ifelse(Legalh > 1, 1, 0),
         LSD_User = ifelse(LSD > 1, 1, 0),
         Meth_User = ifelse(Meth > 1, 1, 0),
         Mushrooms_User = ifelse(Mushrooms > 1, 1, 0),
         Nicotine User = ifelse(Nicotine > 1, 1, 0),
         Semeron_User = ifelse(Semer > 1, 1, 0),
         VSA_User = ifelse(VSA > 1, 1, 0)) %>%
  dplyr::select(Alcohol_User, Amphetamine_User, AmylNitrite_User, Benzos_User,
                Caffeine_User, Cannabis_User, Chocolate_User, Cocaine_User,
                Crack_User, Ecstasy_User, Heroine_User, Ketamine_User,
                LegalHighs_User, LSD_User, Meth_User, Mushrooms_User,
                Nicotine_User, Semeron_User, VSA_User)
data.frame(yes = colSums(drug_byuse), no = 1885 - colSums(drug_byuse),
          drug = colnames(drug_byuse)) %>%
  summarise(diff = abs(yes - no), drug = drug) %>%
  arrange(diff) %>%
 head()
##
   diff
                      drug
## 1 349
              Benzos User
## 2 361 LegalHighs_User
## 3 383
            Ecstasy User
## 4 497
           Mushrooms_User
```

```
## 5 511 Cocaine_User
## 6 529 Amphetamine_User
```

c) Analytical Methods

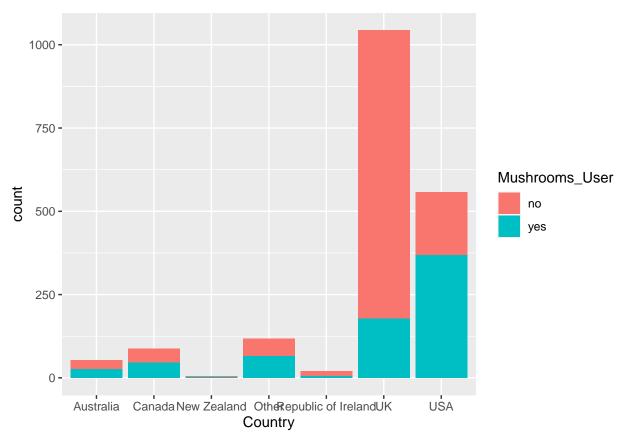


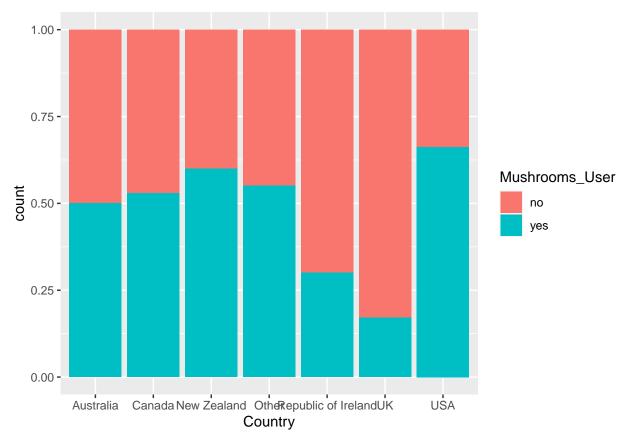
fisher.test(drug_clean_2\$Gender, drug_clean_2\$Mushrooms_User)

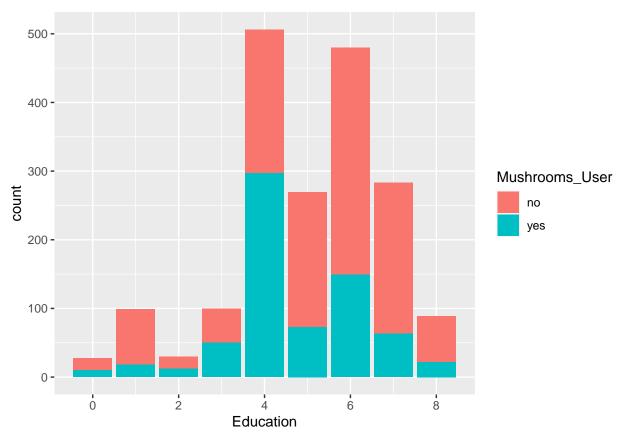
```
##
## Fisher's Exact Test for Count Data
##
## data: drug_clean_2$Gender and drug_clean_2$Mushrooms_User
## p-value < 2.2e-16
## alternative hypothesis: true odds ratio is not equal to 1
## 95 percent confidence interval:
## 2.426426 3.624594
## sample estimates:
## odds ratio
## 2.962888</pre>
```

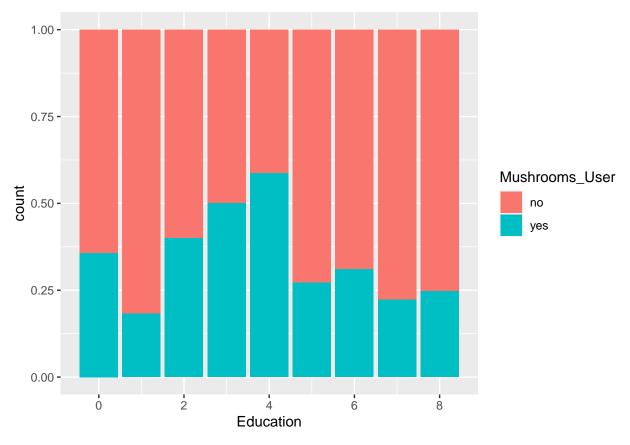
-Since the p-value is less than the significance level, gender is statistically significant.

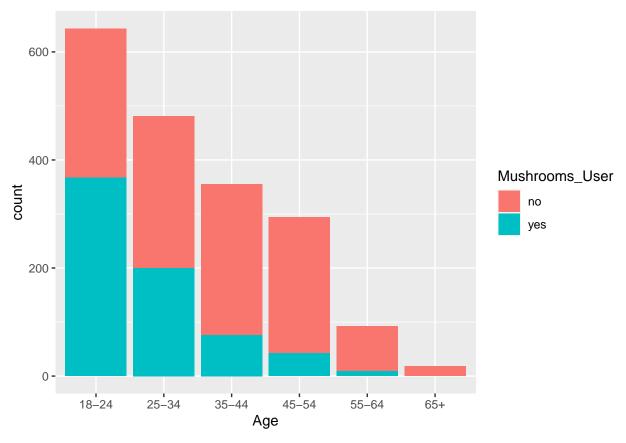


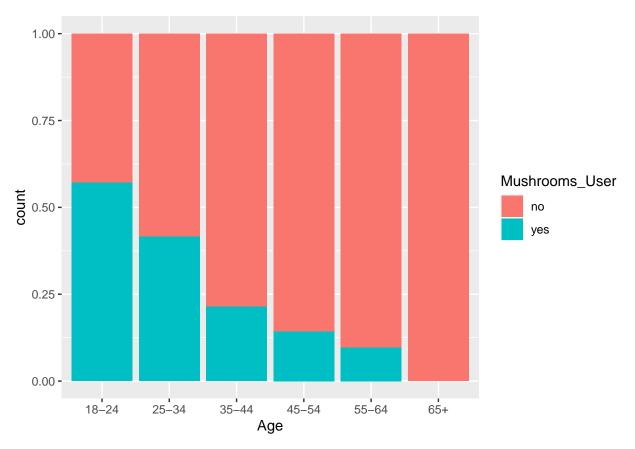


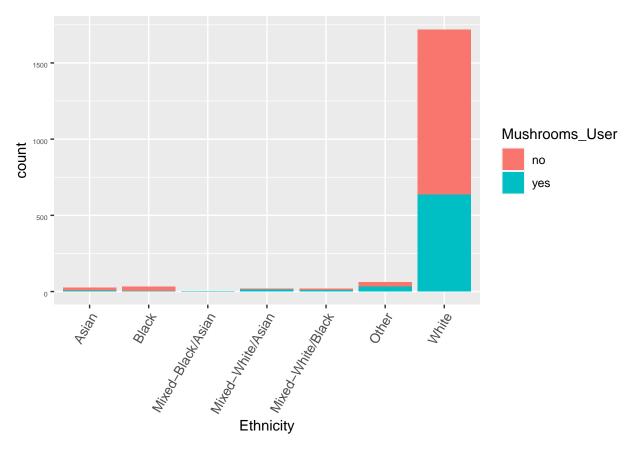


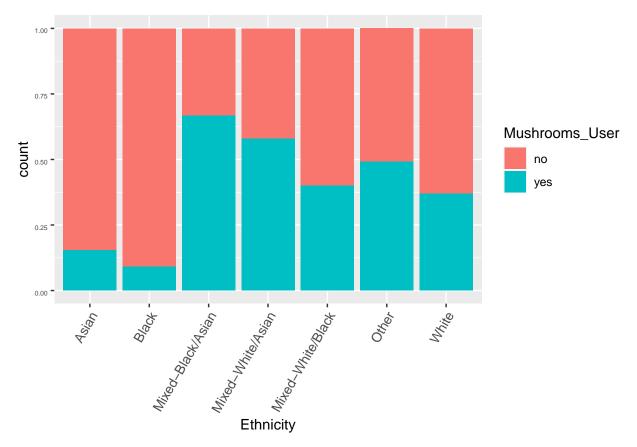




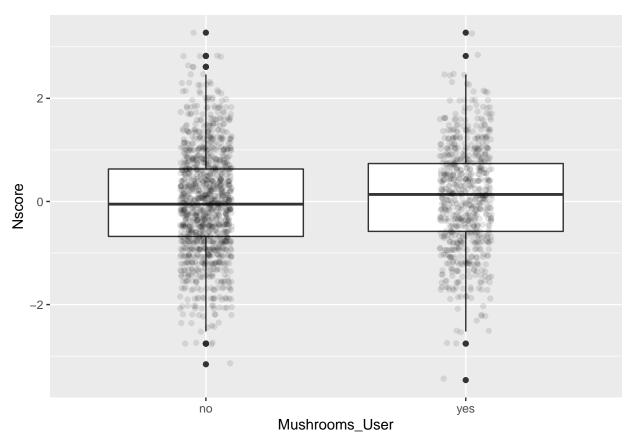




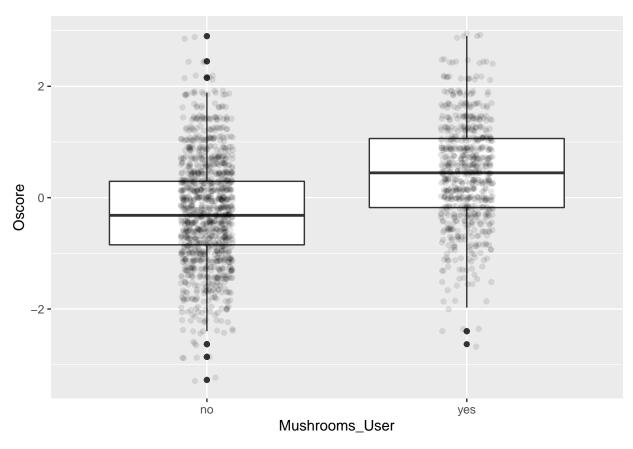




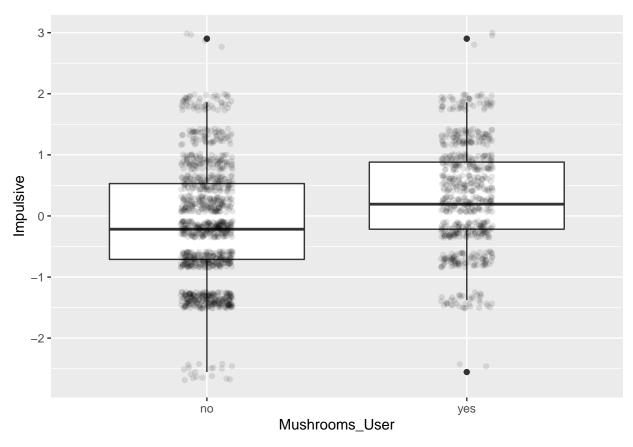
```
drug_clean %>%
  ggplot(aes(Mushrooms_User, Nscore)) +
  geom_boxplot() +
  geom_jitter(width = 0.1, alpha = 0.1)
```



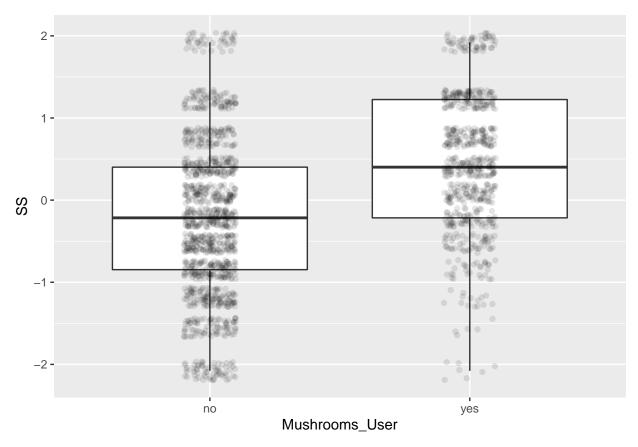
```
drug_clean %>%
  ggplot(aes(Mushrooms_User, Oscore)) +
  geom_boxplot() +
  geom_jitter(width = 0.1, alpha = 0.1)
```



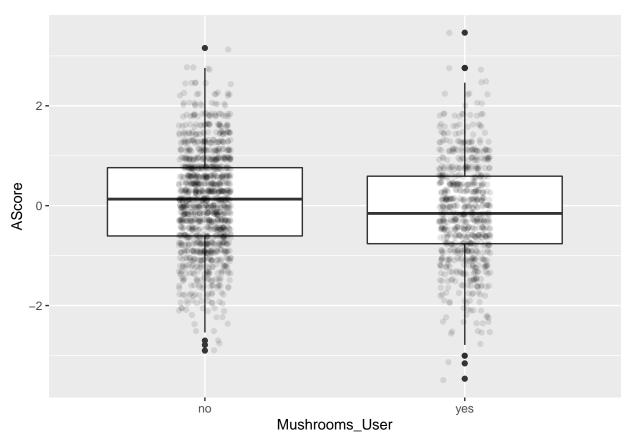
```
drug_clean %>%
  ggplot(aes(Mushrooms_User, Impulsive)) +
  geom_boxplot() +
  geom_jitter(width = 0.1, alpha = 0.1)
```



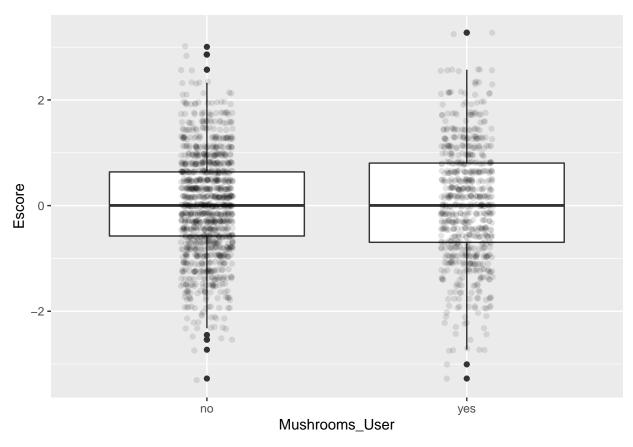
```
drug_clean %>%
  ggplot(aes(Mushrooms_User, SS)) +
  geom_boxplot() +
  geom_jitter(width = 0.1, alpha = 0.1)
```



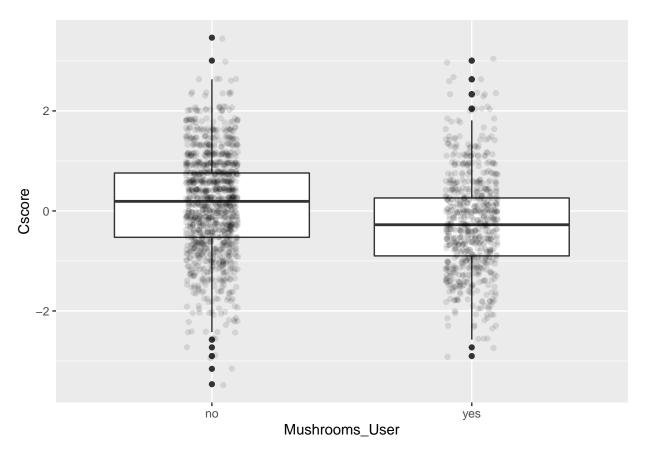
```
drug_clean %>%
   ggplot(aes(Mushrooms_User, AScore)) +
   geom_boxplot() +
   geom_jitter(width = 0.1, alpha = 0.1)
```



```
drug_clean %>%
  ggplot(aes(Mushrooms_User, Escore)) +
  geom_boxplot() +
  geom_jitter(width = 0.1, alpha = 0.1)
```



```
drug_clean %>%
   ggplot(aes(Mushrooms_User, Cscore)) +
   geom_boxplot() +
   geom_jitter(width = 0.1, alpha = 0.1)
```



```
fit_multi <- logistic_reg() %>%
  set_engine("glm") %>%
  fit(Benzos_User ~ Gender + Age + Education + Nscore + Oscore + Impulsive + SS + Cscore + AScore + Esc
result<-tidy(fit_multi, conf.int=TRUE, exponentiate=TRUE)
print(result, n=20)</pre>
```

```
## # A tibble: 22 x 7
##
      term
                            estimate std.error statistic p.value conf.low conf.high
##
      <chr>
                                         <dbl>
                                                             <dbl>
                                                                       <dbl>
                                                                                 <dbl>
                               <dbl>
                                                    <dbl>
                                                                       0.290
                                                                                 0.842
##
    1 (Intercept)
                               0.498
                                        0.271
                                                   -2.58 1.00e- 2
                                                    2.66 7.84e- 3
##
    2 GenderM
                               1.35
                                        0.112
                                                                       1.08
                                                                                 1.68
##
    3 Age25-34
                               1.46
                                        0.145
                                                    2.62 8.83e- 3
                                                                      1.10
                                                                                 1.94
    4 Age35-44
                               1.13
                                        0.162
                                                    0.758 4.48e- 1
                                                                      0.822
                                                                                 1.56
                                                    0.924 3.56e- 1
    5 Age45-54
                                        0.175
                                                                      0.834
                                                                                 1.66
##
                               1.18
##
    6 Age55-64
                               0.674
                                        0.278
                                                   -1.42 1.56e- 1
                                                                      0.384
                                                                                 1.15
                                                   -0.560 5.76e- 1
##
  7 Age65+
                               0.702
                                        0.633
                                                                      0.177
                                                                                 2.23
   8 EducationLeft schoo~
                               1.19
                                        0.332
                                                    0.528 5.98e- 1
                                                                      0.622
                                                                                 2.29
## 9 EducationLeft schoo~
                                                    0.348 7.28e- 1
                                                                      0.470
                                                                                 2.95
                               1.18
                                        0.467
                                                   0.671 5.02e- 1
## 10 EducationLeft schoo~
                                                                      0.652
                                                                                 2.42
                               1.25
                                        0.334
## 11 EducationLeft schoo~
                                                   2.26 2.40e- 2
                                                                                 8.22
                               3.06
                                        0.495
                                                                      1.17
## 12 EducationMasters de~
                                                   -1.29 1.97e- 1
                               0.703
                                        0.273
                                                                      0.413
                                                                                 1.21
## 13 EducationProfession~
                               0.888
                                        0.278
                                                   -0.427 6.69e- 1
                                                                      0.517
                                                                                 1.54
## 14 EducationSome colle~
                               1.23
                                        0.269
                                                   0.755 4.50e- 1
                                                                      0.726
                                                                                 2.09
## 15 EducationUniversity~
                                                   -0.588 5.57e- 1
                                                                      0.519
                               0.859
                                        0.259
                                                                                 1.44
## 16 Nscore
                               1.46
                                        0.0634
                                                    5.97 2.35e- 9
                                                                      1.29
                                                                                 1.66
## 17 Oscore
                                                    6.49 8.48e-11
                               1.49
                                        0.0611
                                                                       1.32
                                                                                 1.68
```

0.989 3.22e- 1 ## 18 Impulsive 1.07 0.0714 0.933 1.23 ## 19 SS 1.30 0.0761 3.49 4.89e- 4 1.12 1.51 ## 20 Cscore -1.19 2.35e- 1 0.929 0.0621 0.822 1.05

... with 2 more rows

Results:

Discussion:

References: