

```

library(readxl)
library(lme4)

## Loading required package: Matrix
library(lmerTest)

##
## Attaching package: 'lmerTest'
## The following object is masked from 'package:lme4':
##
##      lmer
## The following object is masked from 'package:stats':
##
##      step
library(car)

## Loading required package: carData
library(MuMIn)
library(afex)

## *****
## Welcome to afex. For support visit: http://afex.singmann.science/
## - Functions for ANOVAs: aov_car(), aov_ez(), and aov_4()
## - Methods for calculating p-values with mixed(): 'S', 'KR', 'LRT', and 'PB'
## - 'afex_aov' and 'mixed' objects can be passed to emmeans() for follow-up tests
## - NEWS: emmeans() for ANOVA models now uses model = 'multivariate' as default.
## - Get and set global package options with: afex_options()
## - Set orthogonal sum-to-zero contrasts globally: set_sum_contrasts()
## - For example analyses see: browseVignettes("afex")
## *****

##
## Attaching package: 'afex'
## The following object is masked from 'package:lme4':
##
##      lmer
library(ggplot2)

data <- read_excel("../Data/PredictingOutcomes_ParticipantPredictions.xlsx", sheet = "Study 1B")

# divide the data based on the generator
data1 <- data[data$generator == "analyst",]
data2 <- data[data$generator == "bingo",]
data3 <- data[data$generator == "stock",]

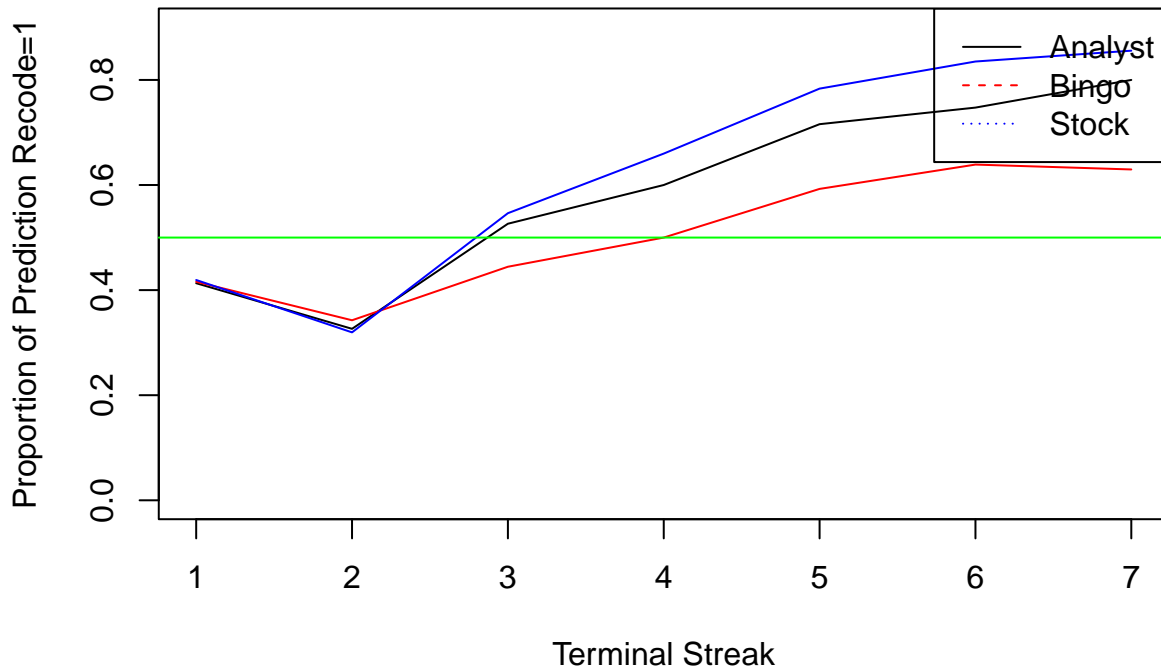
calculate the proportion of participants who predicted the prediction_recode=1 for each terminal_streak_length from 1 to 7

prop1 <- aggregate(data1$prediction_recode, by = list(data1$terminal_streak_length), FUN = mean)
prop2 <- aggregate(data2$prediction_recode, by = list(data2$terminal_streak_length), FUN = mean)
prop3 <- aggregate(data3$prediction_recode, by = list(data3$terminal_streak_length), FUN = mean)

```

```
plot(prop1$Group.1,prop1$x, type = "l",ylim=c(0.0,0.9), xlab = "Terminal Streak", ylab = "Proportion of
lines(prop2$Group.1,prop2$x, col = "red")
lines(prop3$Group.1,prop3$x, col = "blue")
abline(h = 0.5, col = "green")
legend("topright", legend = c("Analyst", "Bingo", "Stock"), col = c("black", "red", "blue"), lty = 1:3)
```

Proportion of Prediction Recode=1 for each Terminal Streak



```
aov1<-aov_ez('participant_id','prediction_recode',data, between=c('generator'),within=c('terminal_streak'))

## Converting to factor: generator

## Warning: More than one observation per design cell, aggregating data using `fun_aggregate = mean`.
## To turn off this warning, pass `fun_aggregate = mean` explicitly.

## Contrasts set to contr.sum for the following variables: generator
aov1

## Anova Table (Type 3 tests)
##
## Response: prediction_recode
##              Effect              df  MSE          F ges p.value
## 1              generator              2, 297 0.50      5.59 ** .014   .004
## 2 terminal_streak_length  5.28, 1567.01 0.15 61.50 *** .114  <.001
## 3 generator:terminal_streak_length 10.55, 1567.01 0.15      1.96 * .008   .031
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '+' 0.1 ' ' 1
##
## Sphericity correction method: GG

pairwise.t.test(data$prediction_recode, data$generator, p.adjust.method = "bonferroni")

##
```

```

## Pairwise comparisons using t tests with pooled SD
##
## data: data$prediction_recode and data$generator
##
##      analyst bingo
## bingo 0.2181  -
## stock 0.7279  0.0079
##
## P value adjustment method: bonferroni
id <- unique(data1$participant_id)

slope <- c()

for (i in id){
  x <- as.character(i)
  datax <- data1[data1$participant_id == x,]
  model <- glm(prediction_recode ~ terminal_streak_length, data = datax, family = binomial)

  beta <- coef(model)[2]
  odds <- exp(beta)
  percent <- (odds-1)*100
  print(percent)
  slope <- c(slope, beta)
}

## terminal_streak_length
##           45.54897
## terminal_streak_length
##           292.5611
## terminal_streak_length
##           15.76128
## terminal_streak_length
##           46.23558

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##           9141762024

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##           -100
## terminal_streak_length
##           98.39988
## terminal_streak_length
##           4.125514
## terminal_streak_length
##           -32.82535
## terminal_streak_length
##           313.0887
## terminal_streak_length
##           -19.7737

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length

```

```

##          13135095123
## terminal_streak_length
##          1.247921
## terminal_streak_length
##          15.76128
## terminal_streak_length
##          -8.700901
## terminal_streak_length
##          189.3349
## terminal_streak_length
##          62.8534
## terminal_streak_length
##          62.8534
## terminal_streak_length
##          -31.72651

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          -100
## terminal_streak_length
##          190.7886
## terminal_streak_length
##          83.60011
## terminal_streak_length
##          -27.21342
## terminal_streak_length
##          49.09466
## terminal_streak_length
##          54.52513
## terminal_streak_length
##          239.9942
## terminal_streak_length
##          19.21552
## terminal_streak_length
##          190.7886
## terminal_streak_length
##          5.344051

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          3607983160

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          9141762012
## terminal_streak_length
##          30.07665

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          6644785868
## terminal_streak_length
##          18.33153
## terminal_streak_length

```

```

##          15.9147
## terminal_streak_length
##          -14.26735

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          9141762026
## terminal_streak_length
##          239.9942
## terminal_streak_length
##          24.6474
## terminal_streak_length
##          134.3743
## terminal_streak_length
##          154.0747
## terminal_streak_length
##          34.6796

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          4906634331

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          4906634327
## terminal_streak_length
##          4.125514
## terminal_streak_length
##          68.58487

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          13135094680
## terminal_streak_length
##          77.32098
## terminal_streak_length
##          22.11936
## terminal_streak_length
##          -13.61533

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          4906634392

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          6644785922

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##          1081748012
## terminal_streak_length
##          85.12092
## terminal_streak_length

```

```

##          124.6156
## terminal_streak_length
##          98.39988
## terminal_streak_length
##          190.7886
## terminal_streak_length
##          775.63

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##          -100

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##          -100

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##          4906634286
## terminal_streak_length
##          190.7886

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##          -100
## terminal_streak_length
##          36.0331
## terminal_streak_length
##          -14.29001
## terminal_streak_length
##          26.10219
## terminal_streak_length
##          134.3743
## terminal_streak_length
##          77.32098
## terminal_streak_length
##          179.937
## terminal_streak_length
##          16.02961
## terminal_streak_length
##          -9.613804
## terminal_streak_length
##          205.2189
## terminal_streak_length
##          124.6156
## terminal_streak_length
##          -7.947992

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
## terminal_streak_length
##          4906634391
## terminal_streak_length
##          239.9942
## terminal_streak_length

```

```

##           179.937
## terminal_streak_length
##           66.4865
## terminal_streak_length
##           239.9942
## terminal_streak_length
##           10.63636

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##           9141762014
## terminal_streak_length
##           48.86569
## terminal_streak_length
##           190.7886

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##           6644785984
## terminal_streak_length
##           99.54388
## terminal_streak_length
##           154.0747

## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##           3607983119
## terminal_streak_length
##           -14.26735
## terminal_streak_length
##           1.247921
## terminal_streak_length
##           -21.15262

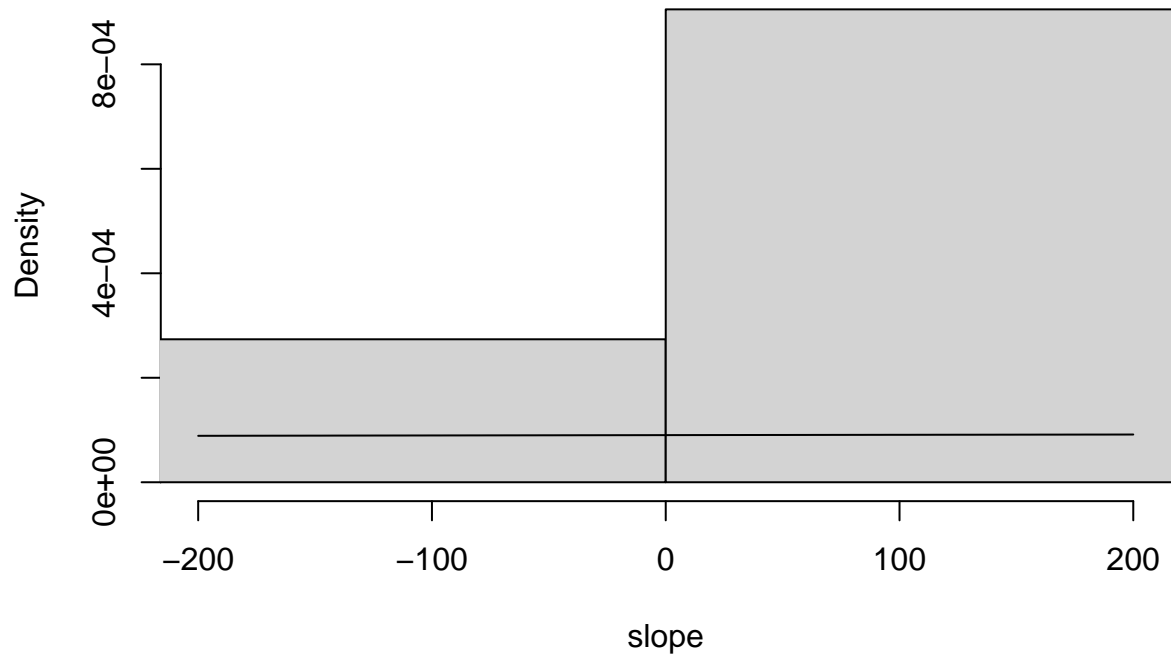
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred

## terminal_streak_length
##           -100
## terminal_streak_length
##           77.32098
## terminal_streak_length
##           105.8674
## terminal_streak_length
##           16.67251
## terminal_streak_length
##           -6.313949

slope <- slope*500
hist(slope,breaks=30,xlim=c(-200,200),prob=TRUE,main="AnalystUnknown")
curve(dnorm(x, mean = mean(slope), sd = sd(slope)), add = TRUE, col = "black")

```

AnalystUnknown



regression

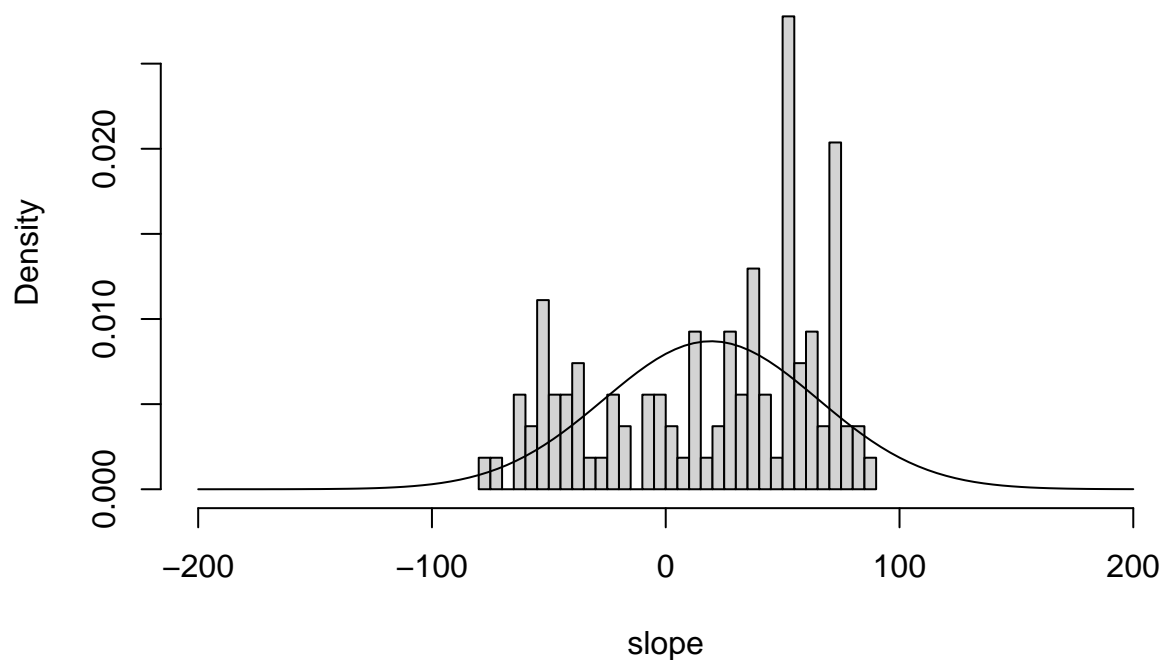
```
id <- unique(data2$participant_id)

slope <- c()

for (i in id){
  x <- as.character(i)
  datax <- data2[data2$participant_id == x,]
  model <- glm(prediction_recode ~ terminal_streak_length, data = datax)
  slope <- c(slope, coef(model)[2])
}
slope <- slope*500

hist(slope,breaks=30,xlim=c(-200,200),prob=TRUE,main="BingoUnknown")
curve(dnorm(x, mean = mean(slope), sd = sd(slope)), add = TRUE, col = "black")
```


BingoUnknown

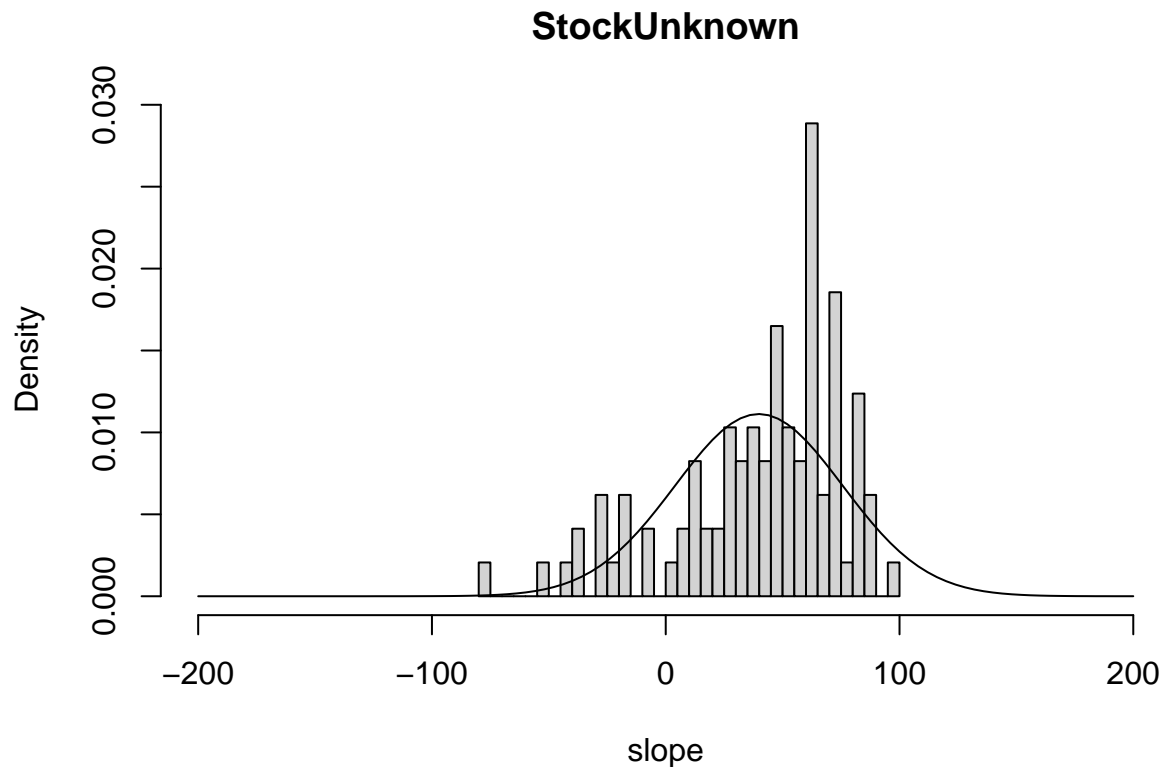


```
id <- unique(data3$participant_id)

slope <- c()

for (i in id){
  x <- as.character(i)
  datax <- data3[data3$participant_id == x,]
  model <- glm(prediction_recode ~ terminal_streak_length, data = datax)
  slope <- c(slope, coef(model)[2])
}
slope <- slope*500

hist(slope,breaks=30,xlim=c(-200,200),prob=TRUE,main="StockUnknown")
curve(dnorm(x, mean = mean(slope), sd = sd(slope)), add = TRUE, col = "black")
```



```
data_dem<- read_excel("../Data/PredictingOutcomes_ParticipantDemographics.xlsx", sheet = "Study 2B")
```

```
# print(data)
```

create a map like data structure to store the unique participant id with there corresponding gender

```
map <- data.frame(unique(data_dem$participant_id), data_dem$gender)
```

```
colnames(map) <- c("participant_id","gender")
```

```
# map
```

```
dataf <- data[,c(2,3,8,10)]
```

```
# print(data1)
```

```
df <- merge(dataf, map, by = "participant_id")
```

```
df_total <- df[df$gender=='0' | df$gender=='1',]
```

```
# male <- df[df$gender=='0',]
```

```
# female <- df[df$gender=='1',]
```

```
# chisq.test(male$prediction_recode, female$prediction_recode,correct=FALSE)
```

```
test <- table(df_total$gender, df_total$prediction_recode)
```

```
print(test)
```

```
##
```

```
##      0      1
```

```
## 0 1467 1341
```

```
## 1 1333 1223
```

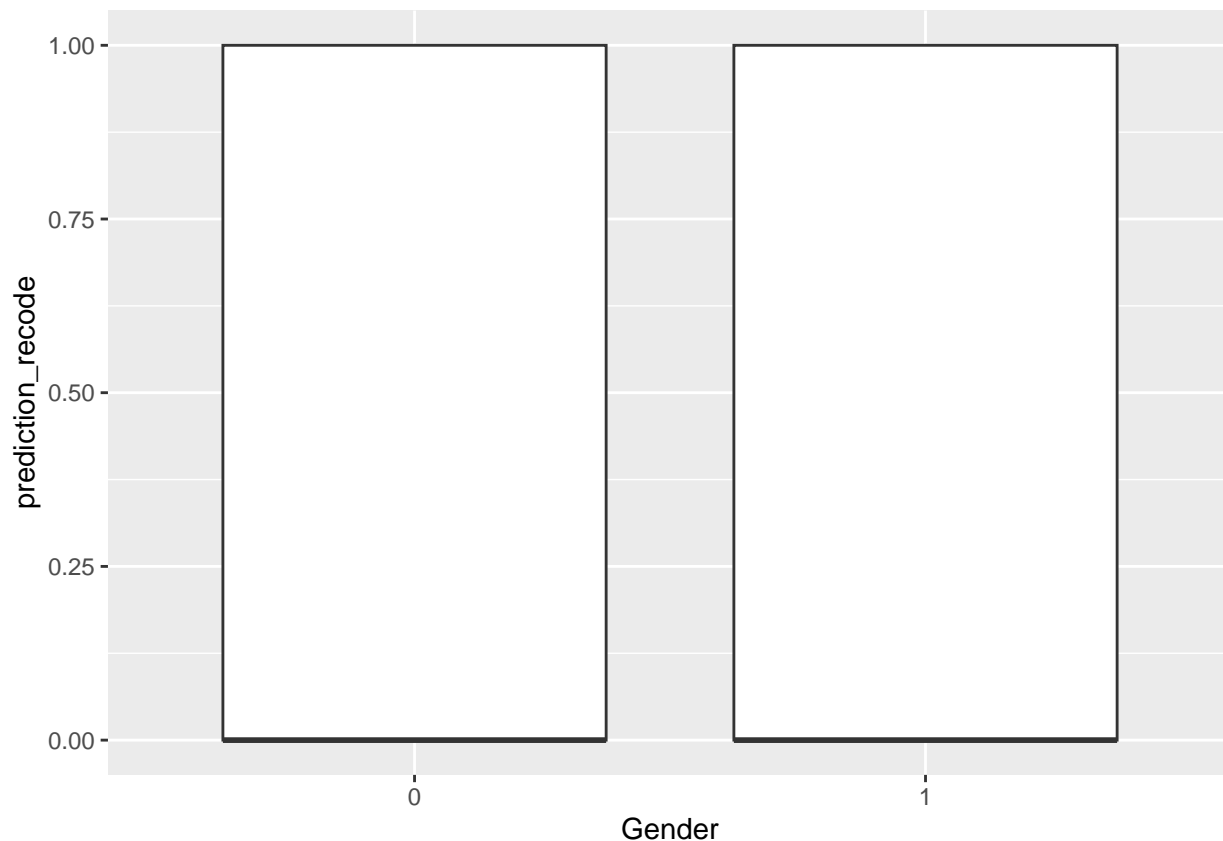
```
chisq.test(test)
```

```
##
```

```
## Pearson's Chi-squared test with Yates' continuity correction
```

```
##
```

```
## data: test
## X-squared = 0.0015883, df = 1, p-value = 0.9682
ggplot(df_total, aes(x = as.factor(gender), y = prediction_recode)) + geom_boxplot() + xlab("Gender")
```



```
df <- df_total[df_total$generator=='analyst',]
test <- table(df$gender, df$prediction_recode)
print(test)
```

```
##
##      0  1
## 0 410 400
## 1 476 424
chisq.test(test)
```

```
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: test
## X-squared = 0.79247, df = 1, p-value = 0.3734
```

```
df <- df_total[df_total$generator=='bingo',]
test <- table(df$gender, df$prediction_recode)
print(test)
```

```
##
##      0  1
## 0 559 449
## 1 485 415
```

```
chisq.test(test)
```

```
##  
## Pearson's Chi-squared test with Yates' continuity correction  
##  
## data: test  
## X-squared = 0.41034, df = 1, p-value = 0.5218  
df <- df_total[df_total$generator=='stock',]  
test <- table(df$gender, df$prediction_recode)  
print(test)
```

```
##  
##      0  1  
## 0 498 492  
## 1 372 384
```

```
chisq.test(test)
```

```
##  
## Pearson's Chi-squared test with Yates' continuity correction  
##  
## data: test  
## X-squared = 0.16469, df = 1, p-value = 0.6849
```

Correlation testing and box-plot form a table with mean of prediction recode on the basis of participant_id and terminal_streak_length

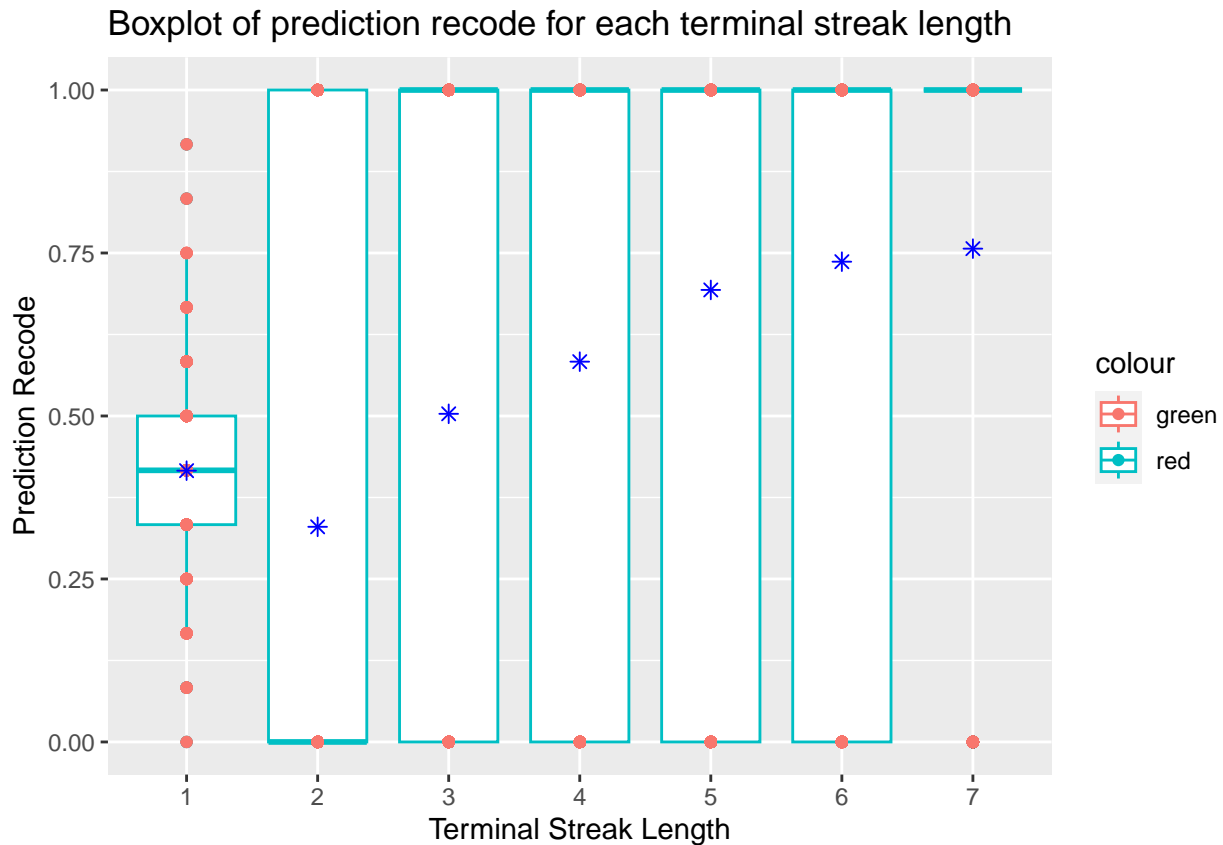
```
cdata <- aggregate(prediction_recode~participant_id+terminal_streak_length,data ,FUN = mean)  
cdata1 <- aggregate(data1$prediction_recode, by = list(data1$participant_id, data1$terminal_streak_length), FUN = mean)  
cdata2 <- aggregate(data2$prediction_recode, by = list(data2$participant_id, data2$terminal_streak_length), FUN = mean)  
cdata3 <- aggregate(data3$prediction_recode, by = list(data3$participant_id, data3$terminal_streak_length), FUN = mean)  
colnames(cdata) <- c( "participant_id","terminal_streak_length","prediction_recode")  
colnames(cdata1) <- c( "participant_id","terminal_streak_length","prediction_recode")  
colnames(cdata2) <- c( "participant_id","terminal_streak_length","prediction_recode")  
colnames(cdata3) <- c( "participant_id","terminal_streak_length","prediction_recode")  
nrow(cdata)
```

```
## [1] 2100
```

```
# cdata
```

draw box plot using ggplot2

```
ggplot(cdata, aes(x = as.factor(terminal_streak_length), y = prediction_recode)) +geom_boxplot(aes(color = terminal_streak_length,  
size = 2, color = "blue")) + ggtitle("Boxplot of prediction recode for each terminal streak length")
```



ply pearson correlation test on the data

```
cor.test(cdata$terminal_streak_length, cdata$prediction_recode, method = "pearson")
```

```
##
## Pearson's product-moment correlation
##
## data: cdata$terminal_streak_length and cdata$prediction_recode
## t = 15.1, df = 2098, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.2739860 0.3511626
## sample estimates:
## cor
## 0.3130911
```

```
cor.test(cdata1$terminal_streak_length, cdata1$prediction_recode, method = "pearson")
```

```
##
## Pearson's product-moment correlation
##
## data: cdata1$terminal_streak_length and cdata1$prediction_recode
## t = 9.3329, df = 663, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.2717788 0.4062701
## sample estimates:
## cor
## 0.3407666
```

```
cor.test(cdata2$terminal_streak_length, cdata2$prediction_recode, method = "pearson")
```

```
##  
## Pearson's product-moment correlation  
##  
## data: cdata2$terminal_streak_length and cdata2$prediction_recode  
## t = 5.9286, df = 754, p-value = 4.65e-09  
## alternative hypothesis: true correlation is not equal to 0  
## 95 percent confidence interval:  
## 0.1418743 0.2781612  
## sample estimates:  
## cor  
## 0.2110432
```

```
cor.test(cdata3$terminal_streak_length, cdata3$prediction_recode, method = "pearson")
```

```
##  
## Pearson's product-moment correlation  
##  
## data: cdata3$terminal_streak_length and cdata3$prediction_recode  
## t = 11.699, df = 677, p-value < 2.2e-16  
## alternative hypothesis: true correlation is not equal to 0  
## 95 percent confidence interval:  
## 0.3455118 0.4708054  
## sample estimates:  
## cor  
## 0.4100916
```