

```

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

print(data)

## # A tibble: 144 x 19
##   study partici~1 gener~2 rate  respo~3 score~4 respo~5 score~6 respo~7 score~8
##   <chr>         <dbl> <chr>   <chr> <chr>         <dbl> <chr>         <dbl> <chr>         <dbl>
## 1 1A             1 analyst ukno~ 50%           0 5             1 0.1%           1
## 2 1A             3 analyst ukno~ 12.5%          1 50             0 0.1%           1
## 3 1A             6 analyst ukno~ 25             0 1             0 1              0
## 4 1A             8 analyst ukno~ 12.5%          1 5             1 0.1%           1
## 5 1A            14 analyst ukno~ 15             0 50             0 0.1            1
## 6 1A            17 analyst ukno~ 50             0 100            0 10             0
## 7 1A            20 analyst ukno~ 15             0 5             1 0.1            1
## 8 1A            24 analyst ukno~ 10             0 5             1 0.1            1

```

```
## 9 1A          28 analyst ukno~ 35          0 5          1 .1          1
## 10 1A         32 analyst ukno~ 0.15%        0 1          0 0.1%        1
## # ... with 134 more rows, 9 more variables: response_fin1 <dbl>,
## #   score_fin1 <dbl>, response_fin2 <dbl>, score_fin2 <dbl>, age <dbl>,
## #   gender <dbl>, highest_degree <dbl>, stocks <dbl>, gambling <dbl>, and
## #   abbreviated variable names 1: participant_id, 2: generator,
## #   3: response_prob1, 4: score_prob1, 5: response_prob2, 6: score_prob2,
## #   7: response_prob3, 8: score_prob3
```

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

```
data1 <- read_excel("../Data/PredictingOutcomes_ParticipantPredictions.xlsx", sheet = "Study 1A")
data_an <- data1[data1$generator == "analyst",]
data_stock <- data1[data1$generator == "stock",]
data_bingo <- data1[data1$generator == "bingo",]
```

make a dataframe with values aggregated on the basis of terminal streak length and participant_id

```
df1 <- aggregate(data1$prediction_recode, by = list(data1$terminal_streak_length, data1$participant_id)
df_an <- aggregate(data_an$prediction_recode, by = list(data_an$terminal_streak_length, data_an$participant_id)
df_stock <- aggregate(data_stock$prediction_recode, by = list(data_stock$terminal_streak_length, data_stock$participant_id)
df_bingo <- aggregate(data_bingo$prediction_recode, by = list(data_bingo$terminal_streak_length, data_bingo$participant_id)
```

change the column name of the dataframe

```
colnames(df1) <- c("terminal_streak_length", "participant_id", "prediction_recode")
colnames(df_an) <- c("terminal_streak_length", "participant_id", "prediction_recode")
colnames(df_stock) <- c("terminal_streak_length", "participant_id", "prediction_recode")
colnames(df_bingo) <- c("terminal_streak_length", "participant_id", "prediction_recode")
shapiro.test(df1$prediction_recode)
```

```
##
## Shapiro-Wilk normality test
##
## data: df1$prediction_recode
## W = 0.95089, p-value < 2.2e-16
```

apply spearman correlation on the dataframe

```
cor.test(df1$terminal_streak_length, df1$prediction_recode, method = "pearson")
```

```
##
## Pearson's product-moment correlation
##
## data: df1$terminal_streak_length and df1$prediction_recode
## t = 16.91, df = 1006, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.4209402 0.5171815
## sample estimates:
## cor
## 0.4704587
```

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

```
##
## Pearson's product-moment correlation
##
## data: df_an$terminal_streak_length and df_an$prediction_recode
```

```
## t = 10.098, df = 348, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.3906956 0.5532504
## sample estimates:
##      cor
## 0.4760289

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

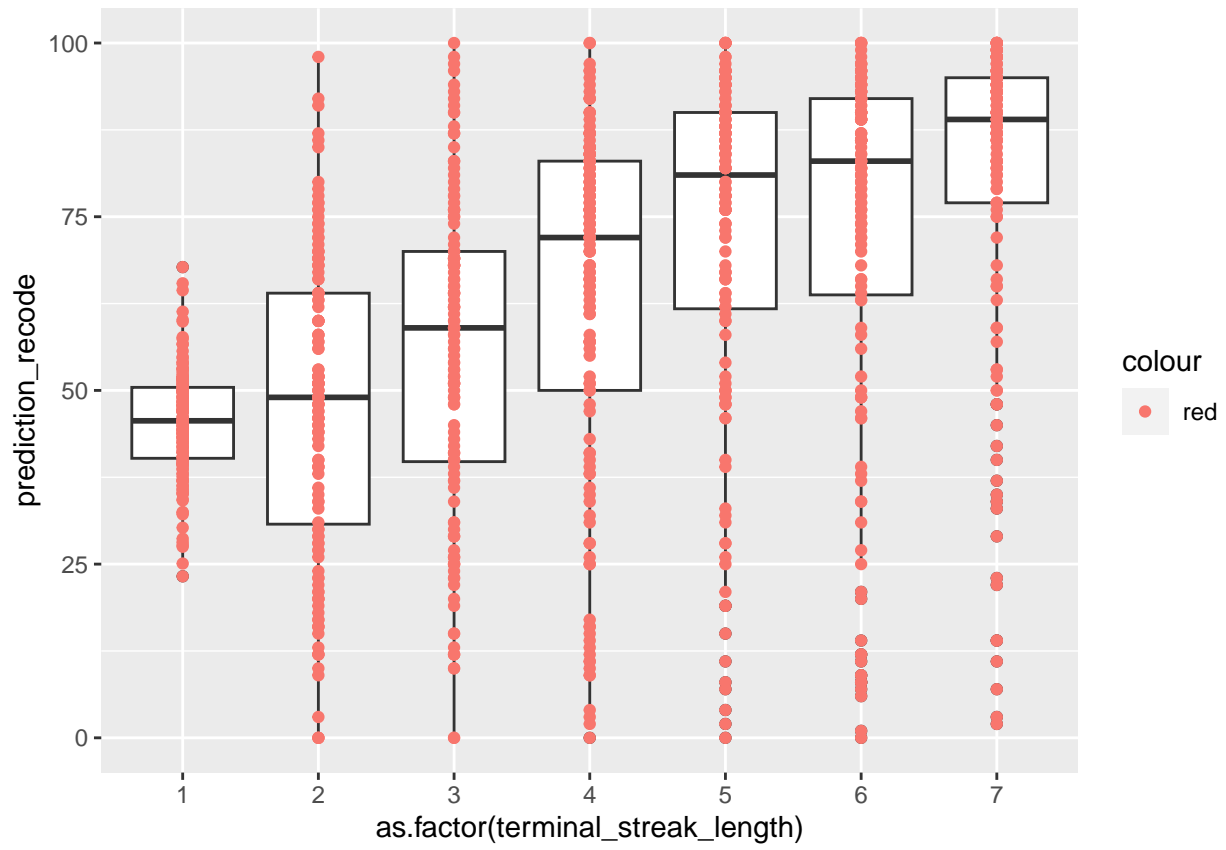
```
##
## Pearson's product-moment correlation
##
## data: df_stock$terminal_streak_length and df_stock$prediction_recode
## t = 12.048, df = 306, p-value < 2.2e-16
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.4862811 0.6384972
## sample estimates:
##      cor
## 0.5672138
```

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

```
##
## Pearson's product-moment correlation
##
## data: df_bingo$terminal_streak_length and df_bingo$prediction_recode
## t = 8.1723, df = 348, p-value = 5.682e-15
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## 0.3094522 0.4856656
## sample estimates:
##      cor
## 0.401265
```

make boxplot of the data having dots with color blue and boxplot with color yellow

```
library(ggplot2)
ggplot(df1, aes(x = as.factor(terminal_streak_length), y = prediction_recode)) + geom_boxplot() + geom_p
```



correlation heatmap for participant_id vs prediction_recode