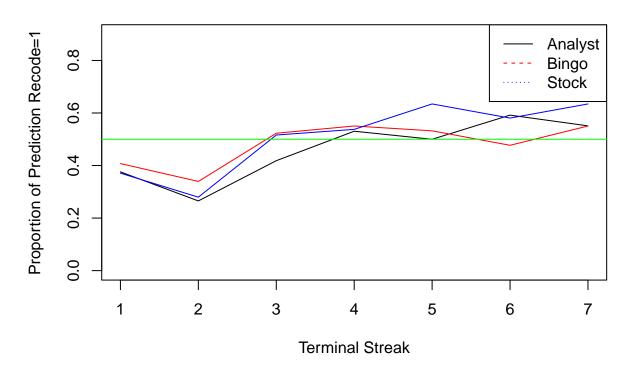
```
library(readxl)
library(afex)
## Loading required package: lme4
## Loading required package: Matrix
## *******
## Welcome to afex. For support visit: http://afex.singmann.science/
## - Functions for ANOVAs: aov_car(), aov_ez(), and aov_4()
## - Methods for calculating p-values with mixed(): 'S', '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_ParticipantPredictions.xlsx", sheet = "Study 3B")
# divide the data based on the generator
data1 <- data[data$generator == "analyst",]</pre>
data2 <- data[data$generator == "bingo",]</pre>
data3 <- data[data$generator == "stock",]</pre>
calculate the proportion of participants who predicted the prediction_recode=1 for each termi-
nal streak length from 1 to 7
prop1 <- aggregate(data1$prediction recode, by = list(data1$terminal streak length), FUN = mean)
prop2 <- aggregate(data2$prediction_recode, by = list(data2$terminal_streak_length), FUN = mean)</pre>
prop3 <- aggregate(data3$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



calculate the effect of condition on participant predicition

```
model1 <- lmer(prediction_recode ~ generator + (1|participant_id), data = data)
summary(model1)</pre>
```

```
## Linear mixed model fit by REML. t-tests use Satterthwaite's method [
## lmerModLmerTest]
## Formula: prediction_recode ~ generator + (1 | participant_id)
##
     Data: data
## REML criterion at convergence: 7666.3
##
## Scaled residuals:
       Min
                1Q Median
                                3Q
                                       Max
## -1.2145 -0.8588 -0.6694 1.0838 1.5428
##
## Random effects:
   Groups
                   Name
                               Variance Std.Dev.
   participant_id (Intercept) 0.01102 0.105
##
                               0.23330 0.483
   Residual
## Number of obs: 5400, groups: participant_id, 300
##
## Fixed effects:
##
                   Estimate Std. Error
                                              df t value Pr(>|t|)
## (Intercept)
                    0.40930
                               0.01564 296.99999
                                                 26.167
                                                            <2e-16 ***
## generatorbingo
                    0.02750
                               0.02156 296.99999
                                                    1.276
                                                             0.203
```

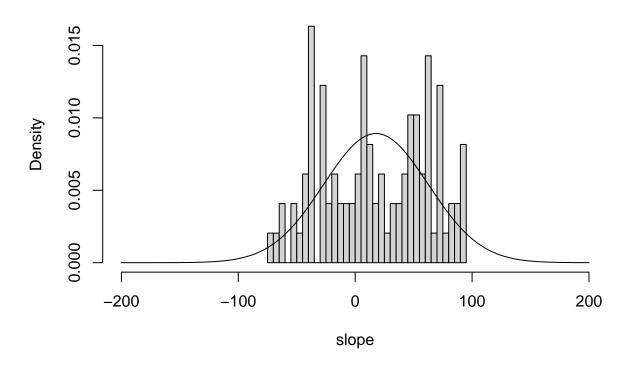
```
## generatorstock 0.01484 0.02242 296.99999 0.662
                                                           0.509
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Correlation of Fixed Effects:
              (Intr) gnrtrb
##
## generatrbng -0.726
## genertrstck -0.698 0.506
anova(model1)
## Type III Analysis of Variance Table with Satterthwaite's method
             Sum Sq Mean Sq NumDF DenDF F value Pr(>F)
## generator 0.37986 0.18993
                                    297 0.8141 0.444
                                2
aov1<-aov_ez('participant_id', 'prediction_recode', data, between=c('generator'), within=c('terminal_strea
## 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
                                                 df MSE
                              Effect
                                                                F ges p.value
## 1
                                             2, 297 0.61
                                                              0.58 .002
                           generator
## 2
              terminal_streak_length 5.16, 1533.11 0.17 23.22 *** .044
## 3 generator:terminal_streak_length 10.32, 1533.11 0.17
                                                              1.52 .006
                                                                           .123
## 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 0.27
## stock 1.00
                1.00
## 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)
    slope <- c(slope, coef(model)[2])
}
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")</pre>
```

AnalystUnknown

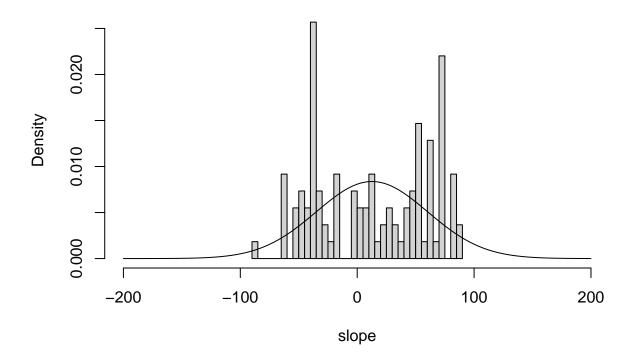


```
id <- unique(data2$participant_id)
slope <- c()

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])
}</pre>
```

```
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")</pre>
```

BingoUnknown



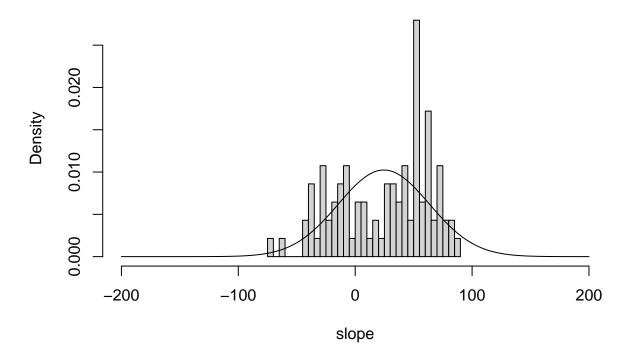
```
id <- unique(data3$participant_id)

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")</pre>
```

StockUnknown



```
data_dem<- read_excel("../Data/PredictingOutcomes_ParticipantDemographics.xlsx", sheet = "Study 2B")
# print(data)</pre>
```

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</pre>
```

```
dataf <- data[,c(2,3,8,10)]
# print(data1)</pre>
```

```
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)
check <- table(df_total$gender, df_total$terminal_streak_length)
print(check)</pre>
```

```
##
##
                      3
                                            7
           1
                2
                           4
                                 5
                                      6
##
     0 1860
              155
                   155
                         155
                               155
                                    155
                                          155
##
     1 1716
             143
                   143
                         143
                               143
                                         143
                                    143
```

```
test <- table(df_total$gender, df_total$prediction_recode)</pre>
print(test)
##
##
          0
               1
     0 1631 1159
##
     1 1458 1116
##
chisq.test(test)
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: test
## X-squared = 1.733, df = 1, p-value = 0.188
df <- df_total[df_total$generator=='analyst',]</pre>
test <- table(df$gender, df$prediction_recode)</pre>
print(test)
##
##
         0
     0 462 330
##
     1 580 392
chisq.test(test)
##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: test
## X-squared = 0.26993, df = 1, p-value = 0.6034
df <- df_total[df_total$generator=='bingo',]</pre>
test <- table(df$gender, df$prediction_recode)</pre>
print(test)
##
##
         0
##
     0 630 468
     1 462 384
chisq.test(test)
##
## Pearson's Chi-squared test with Yates' continuity correction
## data: test
## X-squared = 1.3759, df = 1, p-value = 0.2408
```

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

##
## 0 1
## 0 539 361
## 1 416 340

chisq.test(test)

##
## Pearson's Chi-squared test with Yates' continuity correction
##
## data: test
## X-squared = 3.7826, df = 1, p-value = 0.05179</pre>
```