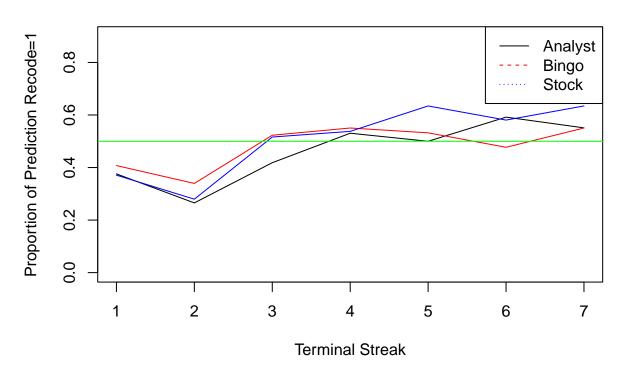
```
library(readxl)
library(afex)
## Loading required package: lme4
## Loading required package: Matrix
## *******
## Welcome to afex. For support visit: http://afex.singmann.science/
## - Functions for ANOVAs: aov_car(), aov_ez(), and aov_4()
## - Methods for calculating p-values with mixed(): 'S', '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")</pre>
# 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>
give count of entries in all three data
nrow(data1)
## [1] 1764
nrow(data2)
## [1] 1962
nrow(data3)
## [1] 1674
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)</pre>
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)
prop1
##
     Group.1
## 1
           1 0.3758503
## 2
           2 0.2653061
           3 0.4183673
## 3
## 4
           4 0.5306122
## 5
           5 0.5000000
## 6
           6 0.5918367
           7 0.5510204
## 7
prop2
     Group.1
##
                     Х
## 1
           1 0.4074924
## 2
           2 0.3394495
## 3
           3 0.5229358
## 4
           4 0.5504587
## 5
           5 0.5321101
## 6
           6 0.4770642
## 7
           7 0.5504587
prop3
##
     Group.1
           1 0.3709677
## 1
## 2
           2 0.2795699
## 3
           3 0.5161290
## 4
           4 0.5376344
## 5
           5 0.6344086
## 6
           6 0.5806452
## 7
           7 0.6344086
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
```

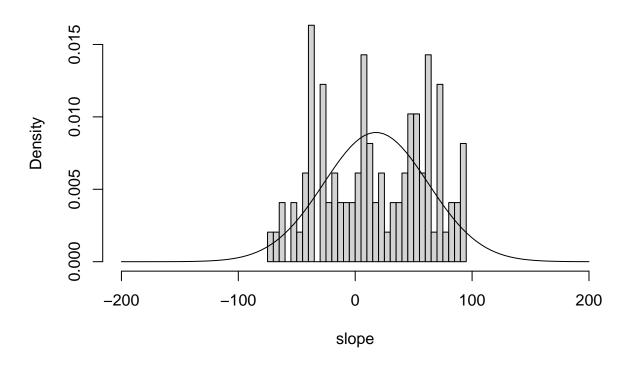
```
## generatorstock 0.01484 0.02242 296.99999 0.662
                                                           0.509
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Correlation of Fixed Effects:
              (Intr) gnrtrb
##
## generatrbng -0.726
## genertrstck -0.698 0.506
anova(model1)
## 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
                           generator
                                             2, 297 0.61
                                                              0.58 .002
## 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

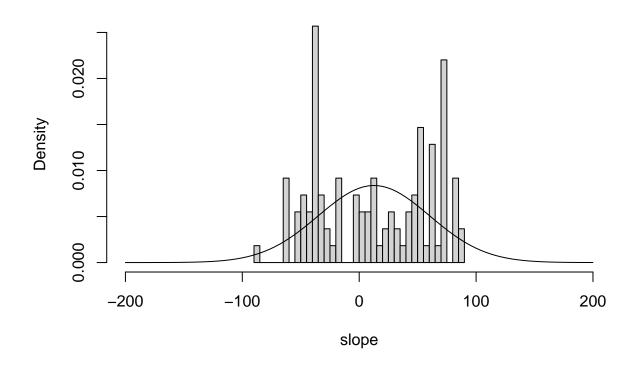


```
id <- unique(data2$participant_id)
slope <- c()

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

