

# Supplemental: Data Analyses for Manuscript

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## Study 1: Age-Related Differences in Motivational Integration and Cognitive Control: Liquid Feedback as Reinforcement

This is an analysis of subject task performance in study 1 in the motivation aging study, in which we examine age-related differences in how individuals integrate the motivational value of monetary and liquid studies, and whether and how this integrated motivational value interacts with cognitive control. In the study, subjects perform a cued task-switching paradigm, where they have to classify a letter (vowel/consonant) or digit (odd/even). During the baseline runs, they don't receive any feedback and are told to respond "as quickly and accurately as possible." During the incentive runs, they receive a drop of liquid (2 mL) to indicate successful attainment of monetary reward in a single trial. Each subject performs 3 incentive runs in which they must respond accurately and faster than a criterion RT to earn monetary reward. Liquid feedback (blocked) served as reinforcement and signaled successful attainment of monetary reward. Criterion RT is calculated on subject performance during the baseline run (30%).

All relevant experimental scripts, subject data, and data analyses located in the online repository in Open Science Framework: <https://osf.io/3mztb/>

### Loading relevant packages

```
library(ggplot2)
library(gridExtra)
library(dplyr)
library(tidyr)
library(broom)
library(lme4)
library(lmerTest)
library(sjPlot)
library(pander)
source("summarySEwithin2.R")
library(RColorBrewer)
```

### Specify colors

```
display.brewer.pal(n = 8, name = 'Set2')
```



Set2 (qualitative)

```
brewer.pal(n = 8, name = "Set2")

## [1] "#66C2A5" "#FC8D62" "#8DA0CB" "#E78AC3" "#A6D854" "#FFD92F" "#E5C494"
## [8] "#B3B3B3"

colors.OAYA<-c("#66C2A5", "#FC8D62")
```

## Path directories of data input/output

```
# data directories
data.path<-paste0("/Users/debbieyee/Box Sync/CCPLab_Liquid_Studies_ARCHIVE/",
                  "2018_MotivationAging/Study1_OA_Reinforcement/SubjectData/",
                  "TrimmedData/OAYA_LiqPos_RewardData.csv")
data.path.base<-paste0("/Users/debbieyee/Box Sync/CCPLab_Liquid_Studies_ARCHIVE/",
                       "2018_MotivationAging/Study1_OA_Reinforcement/SubjectData/",
                       "TrimmedData/OAYA_LiqPos_BaselineData.csv")
data.path.selfreport<-paste0("/Users/debbieyee/Box Sync/CCPLab_Liquid_Studies_ARCHIVE/",
                              "2018_MotivationAging/Study1_OA_Reinforcement/SubjectData/Redcap/",
                              "OAYA_selfreport.csv")
runkey.path<-paste0("/Users/debbieyee/Box Sync/CCPLab_Liquid_Studies_ARCHIVE/",
                    "2018_MotivationAging/Study1_OA_Reinforcement/SubjectData/",
                    "TrimmedData/OAYA_LiqPos_runkey.csv")
figure.path<-paste0("/Users/debbieyee/Box Sync/CCPLab_Liquid_Studies_ARCHIVE/",
                    "2018_MotivationAging/MotivationAging_Analyses/Figures")
```

## Format data for analysis

```
# Read in the data
data<-read.csv(data.path,header = TRUE) %>% select(-X)
```

```

data.base<-read.csv(data.path.base, header = TRUE) %>% select(-X)
runkey<-read.csv(runkey.path, header = TRUE) %>% filter(include=="yes")

# Formatting incentive data
incentive<- data %>% inner_join(y = runkey, by = c("subID","block","group")) %>%
  mutate(subRewarded=(RT<critRT & ACC==1)*1,
    ERR=ifelse(ACC==0,1,0),
    condition="incentive",
    block=factor(block, levels=c(7,8,9), labels=c(1,2,3)),
    money=factor(rewType, levels=c("Reward1","Reward2","Reward4"), labels=c("$","$$","$$$")),
    moneyCode=factor(rewType, levels=c("Reward1","Reward2","Reward4"), labels=c(-1,0,1)),
    moneyweight=factor(rewType, levels=c("Reward1","Reward2","Reward4"), labels=c(.25,.50,1)),
    liqCode=factor(liquid, levels=c("saltwater","neutral","juice"), labels=c(-1,0,1)),
    liqCodeJvN=factor(liquid, levels=c("saltwater","neutral","juice"), labels=c(0,-1,1)),
    liqCodeSvN=factor(liquid, levels=c("saltwater","neutral","juice"), labels=c(-1,1,0)),
    groupCode=factor(group, levels=c("YA","OA"), labels=c(0,1)),
    blockCode=factor(block, levels=c(1,2,3), labels=c(-1,0,1)))
incentive$moneyCode<-as.numeric(levels(incentive$moneyCode)[incentive$moneyCode])
incentive$moneyweight<-as.numeric(levels(incentive$moneyweight)[incentive$moneyweight])
incentive$liqCode<-as.numeric(levels(incentive$liqCode)[incentive$liqCode])
incentive$liqCodeJvN<-as.numeric(levels(incentive$liqCodeJvN)[incentive$liqCodeJvN])
incentive$liqCodeSvN<-as.numeric(levels(incentive$liqCodeSvN)[incentive$liqCodeSvN])
incentive$groupCode<-as.numeric(levels(incentive$groupCode)[incentive$groupCode])
incentive$blockCode<-as.numeric(levels(incentive$blockCode)[incentive$blockCode])

# Calculate average monetary earnings per group
test<-incentive %>% filter(group=="YA") %>% group_by(subID) %>%
  summarise(totalreward=sum(moneyearned))
mean(test$totalreward)

## [1] 4.499444
sd(test$totalreward)

## [1] 0.7665148
test<-incentive %>% filter(group=="OA") %>% group_by(subID) %>%
  summarise(totalreward=sum(moneyearned))
mean(test$totalreward)

## [1] 3.940227
sd(test$totalreward)

## [1] 0.7635672

```

## Summarise/consolidate incentive data

```

# summarise means by subject
groupRT.means = incentive %>% group_by(subID, group) %>%
  filter(!is.na(subRewarded), ACC==1) %>% filter(!is.na(subRewarded), ACC==1) %>%
  summarise(n=n(), meanRT = mean(RT))
group.means = incentive %>% group_by(subID, group) %>% filter(!is.na(subRewarded)) %>%
  summarise(n=n(), meanRR = mean(subRewarded), seRR= sd(subRewarded)/sqrt(length(subRewarded)),
    meanACC = mean(ACC), meanERR = mean(ERR))

```

```

# summarise means by monetary reward only
rewRT.means = incentive %>% group_by(subID, money, group) %>%
  filter(!is.na(subRewarded), ACC==1) %>% filter(!is.na(subRewarded), ACC==1) %>%
  summarise(n=n(), meanRT = mean(RT))
rew.means = incentive %>% group_by(subID, money, group) %>% filter(!is.na(subRewarded)) %>%
  summarise(n=n(), meanRR = mean(subRewarded), meanACC = mean(ACC), meanERR = mean(ERR))

# summarise means by liquid incentive type only
liqRT.means = incentive %>% group_by(subID, liquid, group, block) %>%
  filter(!is.na(subRewarded), ACC==1) %>%
  summarise(meanRT = mean(RT))
liq.means=incentive %>% group_by(subID, liquid, group, block) %>% filter(!is.na(subRewarded)) %>%
  summarise(n=n(), meanRR = mean(subRewarded), meanACC = mean(ACC), meanERR = mean(ERR))

# summarise the means of the incentive data for each subject, grouped by condition
incentive9RT.means <- incentive %>% group_by(subID, liquid, money, group, block) %>%
  filter(!is.na(subRewarded), ACC==1) %>% summarise(n=n(), meanRT = mean(RT)) %>%
  mutate(moneycode = as.numeric(as.character(factor(money, levels=c("$", "$$", "$$$$"),
    labels=c(-1,0,1)))),
    liqcode = as.numeric(as.character(factor(liquid, levels=c("saltwater", "neutral", "juice"),
    labels=c(-1,0,1))))) %>%
  ungroup(subID) %>% mutate(subID=as.factor(subID))
incentive9.means = incentive %>% group_by(subID, liquid, money, group, block) %>%
  filter(!is.na(subRewarded)) %>%
  summarise(n = n(), meanRR = mean(subRewarded), meanACC = mean(ACC), meanERR = mean(ERR)) %>%
  mutate(moneycode = as.numeric(as.character(factor(money, levels=c("$", "$$", "$$$$"),
    labels=c(-1,0,1)))),
    liqcode = as.numeric(as.character(factor(liquid, levels=c("saltwater", "neutral", "juice"),
    labels=c(-1,0,1))))) %>%
  ungroup(subID) %>% mutate(subID=as.factor(subID))

```

# Monetary Rewards Improve Cognitive Task Performance for Both Older and Younger Adults

Plot: histogram of reward rate by age group

```
RR.sum=summarySE2(data=group.means, measurevar = "meanRR", groupvars = c("group"))
```

```
##
## Attaching package: 'data.table'
## The following objects are masked from 'package:dplyr':
##
##   between, first, last
```

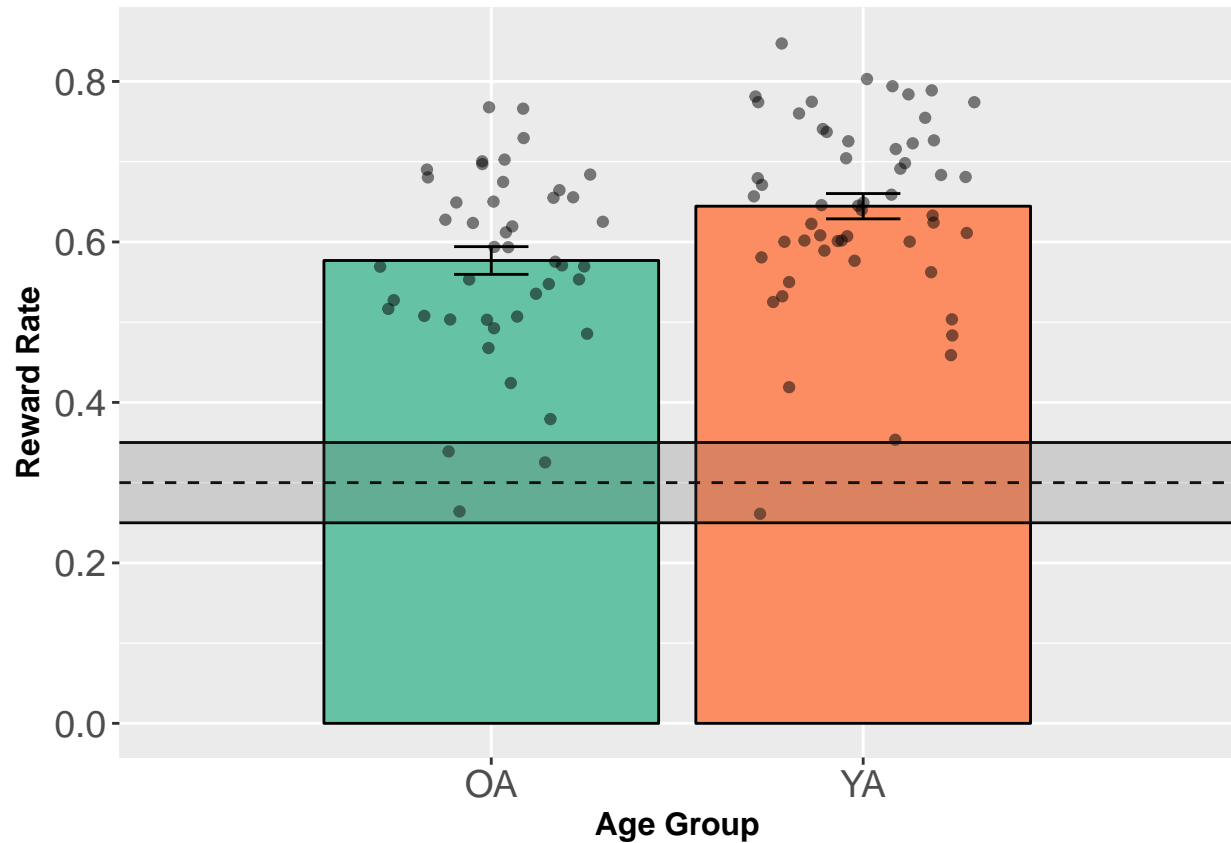
```
pandoc.table(RR.sum)
```

```
##
## -----
##  group    N   meanRR    sd      se      ci
## -----
##   OA     44   0.5769   0.1146  0.01727  0.03483
##
##   YA     54   0.6445   0.1159  0.01577  0.03162
## -----
```

```
p.RR.0<-ggplot() +
  geom_bar(data = RR.sum, aes(x=group, y=meanRR, fill=group),
    color="black",stat="identity") +
  geom_errorbar(data = RR.sum,
    mapping=aes(x=group, y=meanRR,
      ymin=meanRR-se, ymax=meanRR+se),
    stat="identity", width=0.2) +
  geom_point(data = group.means, aes(x=group, y=meanRR),
    stat="identity", position=position_jitter(w=.3), alpha=0.5) +
  geom_hline(yintercept = .30, linetype="dashed") +
  geom_hline(yintercept = .35) +
  geom_hline(yintercept = .25) +
  annotate("rect", xmin = 0, xmax = 3, ymin = .25, ymax = .35,
    alpha = .2) +
  xlab("Age Group") + ylab("Reward Rate") +
  coord_cartesian(ylim=c(0,.85)) +
  scale_fill_brewer(palette="Set2") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
    axis.title=element_text(size=12,face = "bold"),
    axis.text=element_text(size=14),
    legend.position="none",
    strip.text.x = element_text(size = 12))
```

```
## Warning: Ignoring unknown aesthetics: y
```

p.RR.0



## Binomial test

```
binom.test(x = 86, n = 288, p = .05, alternative = "two.sided")
```

```
##
## Exact binomial test
##
## data: 86 and 288
## number of successes = 86, number of trials = 288, p-value <
## 2.2e-16
## alternative hypothesis: true probability of success is not equal to 0.05
## 95 percent confidence interval:
## 0.2463360 0.3550857
## sample estimates:
## probability of success
## 0.2986111
```

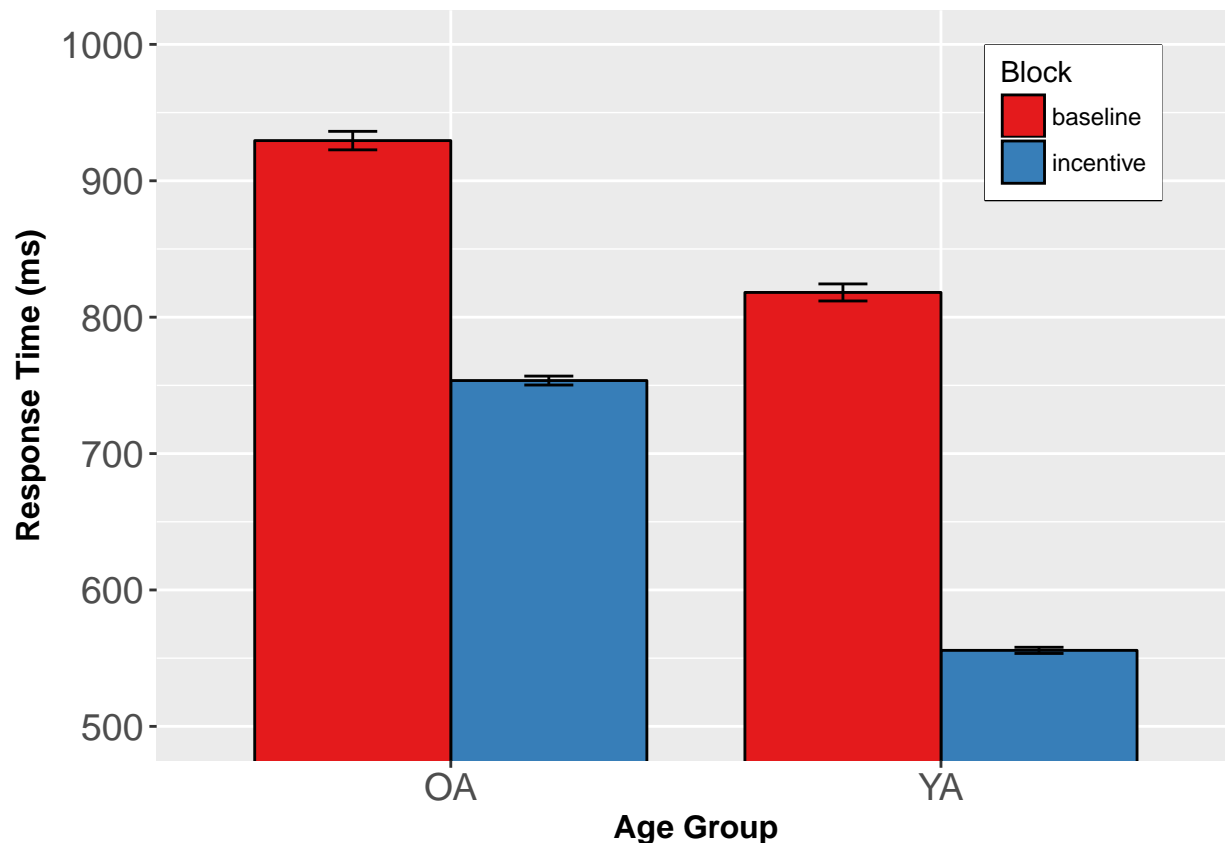
Plot: RT between baseline and incentive blocks.

```
# Merge baseline and incentive data
cond.RT.base <- data.base %>%
  mutate(block=0) %>%
  select(subID, group, Trial, RT, ACC, condition, block, taskswitch=taskSwitch) %>%
  filter(ACC==1)
cond.RT.rew <- incentive %>%
  select(subID, group, Trial, RT, ACC, condition, block, taskswitch) %>%
  filter(ACC==1)
cond.RT<-rbind(cond.RT.base,cond.RT.rew)

RT.sum=summarySEwithin2(data=cond.RT, measurevar = "RT", withinvars = c("condition"),
                        betweenvars = c("group"), idvar = "subID")
pandoc.table(RT.sum)
```

```
##
## -----
##  group    condition      N    RT    RTNormed    sd    se    ci
## -----
##   OA      baseline    3703   929.5    842.9    412.7   6.782   13.3
##
##   OA      incentive   10983   753.5    659.2    342.4   3.267   6.403
##
##   YA      baseline    4620   818.1    897.4    424.6   6.247   12.25
##
##   YA      incentive   12525   555.7    634.8    254.3   2.272   4.454
## -----
```

```
p.RT.0<-ggplot(data = RT.sum, aes(x=group, y = RT, fill=condition)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
           stat="identity", width=0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
               aes(ymin=RT-se, ymax=RT+se), width=.2) +
  xlab("Age Group") + ylab("Response Time (ms)") +
  labs(fill = "Block") +
  coord_cartesian(ylim=c(500,1000)) +
  scale_fill_brewer(palette="Set1") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position=c(.85,.85),
        legend.box.background = element_rect(colour = "black"),
        strip.text.x = element_text(size = 12))
p.RT.0
```



RT by baseline vs incentive block for each age group

```
# consolidating RT data between baseline and incentive
cond.RT<-cond.RT %>%
  mutate(groupCode=factor(group, levels=c("YA","OA"), labels=c(0,1)),
         conditionCode=factor(condition, levels=c("baseline","incentive"), labels=c(1,0)))
cond.RT$groupCode<-as.numeric(levels(cond.RT$groupCode)[cond.RT$groupCode])
cond.RT$conditionCode<-as.numeric(levels(cond.RT$conditionCode)[cond.RT$conditionCode])

# summarizing to have one datapoint per subject per condition.
cond.RT.sum <- cond.RT %>% group_by(subID,group,condition) %>%
  summarise(meanRT = mean(RT))

# OA baseline vs incentive block
t.test(x = subset(cond.RT.sum, group=="OA" & condition=="baseline")$meanRT,
       y = subset(cond.RT.sum, group=="OA" & condition=="incentive")$meanRT,
       paired = TRUE)
```

```
##
## Paired t-test
##
## data: subset(cond.RT.sum, group == "OA" & condition == "baseline")$meanRT and subset(cond.RT.sum, g
## t = 11.035, df = 43, p-value = 4.001e-14
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
```



```
## 156.1623 226.0082
## sample estimates:
## mean of the differences
## 191.0853
# YA baseline vs incentive block
t.test(x = subset(cond.RT.sum, group=="YA" & condition=="baseline")$meanRT,
       y = subset(cond.RT.sum, group=="YA" & condition=="incentive")$meanRT,
       paired = TRUE)

##
## Paired t-test
##
## data: subset(cond.RT.sum, group == "YA" & condition == "baseline")$meanRT and subset(cond.RT.sum, g
## t = 16.346, df = 53, p-value < 2.2e-16
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 235.1593 300.9423
## sample estimates:
## mean of the differences
## 268.0508
```

## Plot: RT Switch costs between baseline and incentive blocks

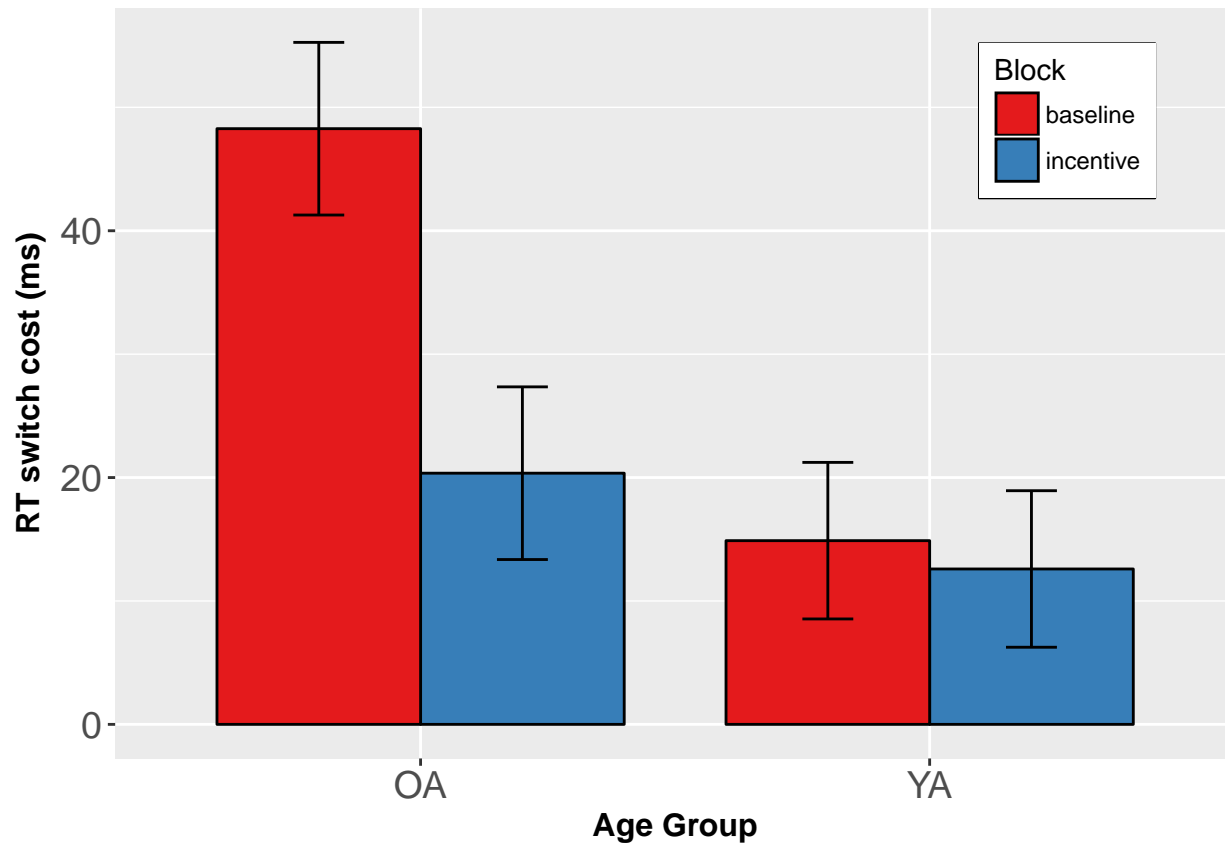
```
cond.RT.switch <- cond.RT %>%
  group_by(subID,group,condition,taskswitch) %>%
  filter(Trial!=1, ACC==1) %>%
  summarise(meanRT = mean(RT)) %>%
  spread(key = taskswitch, value = meanRT) %>%
  rename(taskrepeat="0",taskswitch="1") %>%
  mutate(switchcost = taskswitch-taskrepeat)

SW.sum=summarySEwithin2(data=cond.RT.switch, measurevar = "switchcost",
  withinvars = c("condition"), betweenvars = c("group"),
  idvar = "subID")

pandoc.table(SW.sum)
```

```
##
## -----
##  group    condition    N    switchcost    switchcostNormed    sd    se    ci
## -----
##    OA      baseline    44     48.27          35.47          46.4    6.995    14.11
##
##    OA      incentive    44     20.35          10.48          46.4    6.995    14.11
##
##    YA      baseline    54     14.89          24.12          46.61    6.343    12.72
##
##    YA      incentive    54     12.59          21.83          46.61    6.343    12.72
## -----
```

```
p.SW.1<-ggplot(data = SW.sum, aes(x=group, y = switchcost, fill=condition)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
    stat="identity", width=0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
    aes(ymin=switchcost-se, ymax=switchcost+se), width=.2) +
  xlab("Age Group") + ylab("RT switch cost (ms)") +
  labs(fill = "Block") +
  #coord_cartesian(ylim=c(500,1000)) +
  scale_fill_brewer(palette="Set1") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
    axis.title=element_text(size=12,face = "bold"),
    axis.text=element_text(size=14),
    legend.position=c(.85,.85),
    legend.box.background = element_rect(colour = "black"),
    strip.text.x = element_text(size = 12))
p.SW.1
```



## RT switch costs analysis

```
# OA baseline vs incentive block
t.test(x = subset(cond.RT.switch, group=="OA" & condition=="baseline")$switchcost,
       y = subset(cond.RT.switch, group=="OA" & condition=="incentive")$switchcost,
       paired = TRUE)
```

```
##
## Paired t-test
##
## data: subset(cond.RT.switch, group == "OA" & condition == "baseline")$switchcost and subset(cond.RT
## t = 2.7703, df = 43, p-value = 0.008236
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  7.594471 48.238163
## sample estimates:
## mean of the differences
##          27.91632
```

```
# YA baseline vs incentive block
t.test(x = subset(cond.RT.switch, group=="YA" & condition=="baseline")$switchcost,
       y = subset(cond.RT.switch, group=="YA" & condition=="incentive")$switchcost,
       paired = TRUE)
```

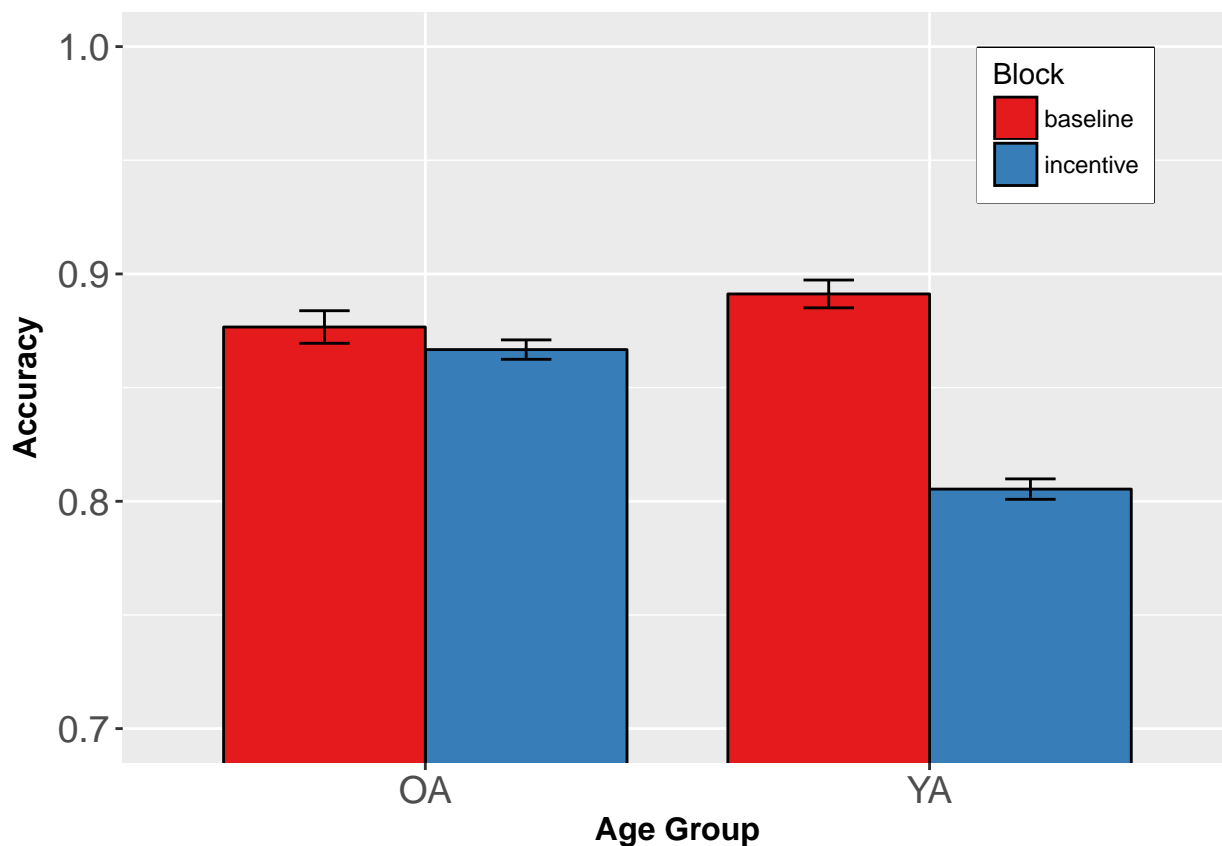
```
##
## Paired t-test
```

```
##
## data:  subset(cond.RT.switch, group == "YA" & condition == "baseline")$switchcost and subset(cond.RT
## t = 0.25574, df = 53, p-value = 0.7991
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -15.69783  20.28578
## sample estimates:
## mean of the differences
##                2.293979
```

## Plot: accuracy between baseline and incentive blocks

```
# Merge baseline and incentive data
cond.ACC.base <- data.base %>% mutate(block=0) %>%
  select(subID, group, Trial, RT, ACC, condition, block)
cond.ACC.rew <- incentive %>% select(subID, group, Trial, RT, ACC, condition, block)
cond.ACC<-rbind(cond.ACC.base,cond.ACC.rew)

ACC.sum=summarySEwithin2(data=cond.ACC, measurevar = "ACC", withinvars = c("condition"),
  betweenvars = c("group"), idvar = "block")
p.ACC.0<-ggplot(data = ACC.sum, aes(x=group, y = ACC, fill=condition)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
    stat="identity", width = 0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
    aes(ymin=ACC-se, ymax=ACC+se), width=.2) +
  xlab("Age Group") + ylab("Accuracy") +
  labs(fill = "Block") +
  coord_cartesian(ylim=c(.7,1)) +
  scale_fill_brewer(palette="Set1") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
    axis.title=element_text(size=12,face = "bold"),
    axis.text=element_text(size=14),
    legend.position=c(.85,.85),
    legend.box.background = element_rect(colour = "black"),
    strip.text.x = element_text(size = 12))
p.ACC.0
```



## Accuracy by baseline vs incentive block for each age group

```
# consolidating accuracy data between baseline and incentive
cond.ACC<-cond.ACC %>%
  mutate(groupCode=factor(group, levels=c("YA","OA"), labels=c(0,1)),
         conditionCode=factor(condition, levels=c("baseline","incentive"), labels=c(1,0)))
cond.ACC$groupCode<-as.numeric(levels(cond.ACC$groupCode)[cond.ACC$groupCode])
cond.ACC$conditionCode<-as.numeric(levels(cond.ACC$conditionCode)[cond.ACC$conditionCode])

# summarizing to have one datapoint per subject per condition.
cond.ACC.sum <- cond.ACC %>% group_by(subID,group,condition) %>%
  summarise(meanACC = mean(ACC))
```

```
# OA baseline vs incentive block
```

```
t.test(x = subset(cond.ACC.sum, group=="OA" & condition=="baseline")$meanACC,
       y = subset(cond.ACC.sum, group=="OA" & condition=="incentive")$meanACC,
       paired = TRUE)
```

```
##
```

```
## Paired t-test
```

```
##
```

```
## data: subset(cond.ACC.sum, group == "OA" & condition == "baseline")$meanACC and subset(cond.ACC.sum
```

```
## t = 0.58738, df = 43, p-value = 0.56
```

```
## alternative hypothesis: true difference in means is not equal to 0
```

```
## 95 percent confidence interval:
```

```
## -0.02419529 0.04408165
```

```
## sample estimates:
```

```
## mean of the differences
```

```
## 0.009943182
```

```
# YA baseline vs incentive block
```

```
t.test(x = subset(cond.ACC.sum, group=="YA" & condition=="baseline")$meanACC,
       y = subset(cond.ACC.sum, group=="YA" & condition=="incentive")$meanACC,
       paired = TRUE)
```

```
##
```

```
## Paired t-test
```

```
##
```

```
## data: subset(cond.ACC.sum, group == "YA" & condition == "baseline")$meanACC and subset(cond.ACC.sum
```

```
## t = 5.0377, df = 53, p-value = 5.814e-06
```

```
## alternative hypothesis: true difference in means is not equal to 0
```

```
## 95 percent confidence interval:
```

```
## 0.05166385 0.12001825
```

```
## sample estimates:
```

```
## mean of the differences
```

```
## 0.08584105
```

## Generalized Linear Mixed Effects Models: Reward Rate Effects by Monetary Reward and Age

```
# Monetary Reward Effect
m.RR.money<-glmer(formula = subRewarded ~ moneyCode * groupCode + (1+moneyCode|subID),
  data = incentive, family = binomial)
summary(m.RR.money)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode * groupCode + (1 + moneyCode | subID)
## Data: incentive
##
##           AIC          BIC    logLik deviance df.resid
## 36155.7 36213.4 -18070.9 36141.7    28217
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.5238 -1.0790  0.6020  0.7781  1.8581
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.23811  0.4880
## moneyCode 0.01212  0.1101  -0.16
## Number of obs: 28224, groups: subID, 98
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.63110    0.06859   9.201 < 2e-16 ***
## moneyCode      0.24920    0.02613   9.536 < 2e-16 ***
## groupCode     -0.30862    0.10215  -3.021 0.00252 **
## moneyCode:groupCode -0.16710    0.03835  -4.358 1.31e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd gropCd
## moneyCode   -0.075
## groupCode   -0.670  0.050
## mnyCd:grpCd  0.051 -0.681 -0.083
```

```
# YA
m.RR.money.YA <-glmer(formula = subRewarded ~ moneyCode + (1+moneyCode|subID),
  data = subset(incentive,groupCode==0), family = binomial)
summary(m.RR.money.YA)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode + (1 + moneyCode | subID)
## Data: subset(incentive, groupCode == 0)
##
##           AIC          BIC    logLik deviance df.resid
```

```

## 19409.8 19448.0 -9699.9 19399.8 15547
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.5666 -1.0924  0.5779  0.7372  1.8434
##
## Random effects:
##      Groups Name      Variance Std.Dev. Corr
## subID (Intercept) 0.25574  0.5057
##      moneyCode  0.01781  0.1335  -0.05
## Number of obs: 15552, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.63220    0.07102   8.902  <2e-16 ***
## moneyCode    0.25050    0.02813   8.904  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr)
## moneyCode -0.019
# OA
m.RR.money.OA <-glmer(formula = subRewarded ~ moneyCode + (1+moneyCode|subID),
                      data = subset(incentive,group=="OA"), family = binomial)
summary(m.RR.money.OA)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode + (1 + moneyCode | subID)
## Data: subset(incentive, group == "OA")
##
##      AIC      BIC    logLik deviance df.resid
## 16749.7 16786.9 -8369.8 16739.7 12667
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.7980 -1.0783  0.6617  0.8247  1.7064
##
## Random effects:
##      Groups Name      Variance Std.Dev. Corr
## subID (Intercept) 0.217087 0.46593
##      moneyCode  0.005865 0.07659  -0.39
## Number of obs: 12672, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.32200    0.07264   4.433 9.29e-06 ***
## moneyCode    0.08181    0.02542   3.218  0.00129 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:

```



```
##          (Intr)
## moneyCode -0.171
```

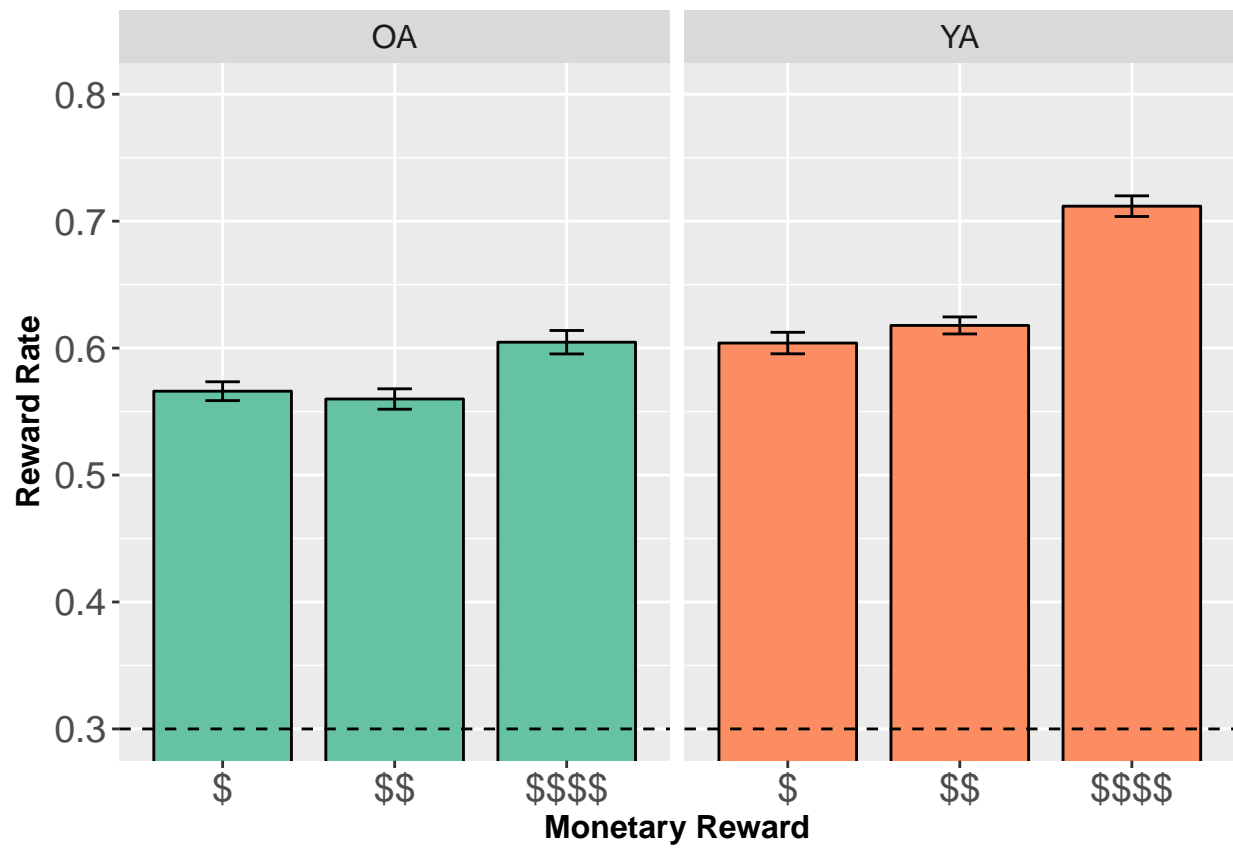
## Plot: barplot of reward rate by monetary reward and age group

```
RR.sum=summarySEwithin2(data=rew.means, measurevar = "meanRR", withinvars = c("money"),
                        betweenvars = c("group"), idvar = "subID")
pandoc.table(RR.sum)
```

```
##
## -----
##  group  money   N   meanRR   meanRRNormed   sd      se      ci
## -----
##   OA      $    44   0.5661     0.6033     0.04918  0.007414  0.01495
##
##   OA     $$    44   0.5599     0.5972     0.05329  0.008034  0.0162
##
##   OA    $$$$    44   0.6046     0.6419     0.06127  0.009237  0.01863
##
##   YA      $    54   0.604     0.5736     0.06216  0.008459  0.01697
##
##   YA     $$    54   0.6179     0.5875     0.04929  0.006707  0.01345
##
##   YA    $$$$    54   0.7118     0.6814     0.05975  0.008131  0.01631
## -----
```

```
p.RR.1<-ggplot(RR.sum, aes(x=money, y=meanRR,fill=group)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
           stat="identity", width=0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
               aes(ymin=meanRR-se, ymax=meanRR+se), width=.2) +
  facet_grid(.~group) +
  geom_hline(yintercept = .30, linetype="dashed") +
  xlab("Monetary Reward") + ylab("Reward Rate") +
  #ggtitle("Task Performance Across Money") +
  coord_cartesian(ylim=c(.3,.8)) +
  #scale_fill_discrete(name="Monetary Reward") +
  scale_fill_brewer(palette="Set2") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position="none",
        strip.text.x = element_text(size = 12))
```

```
p.RR.1
```



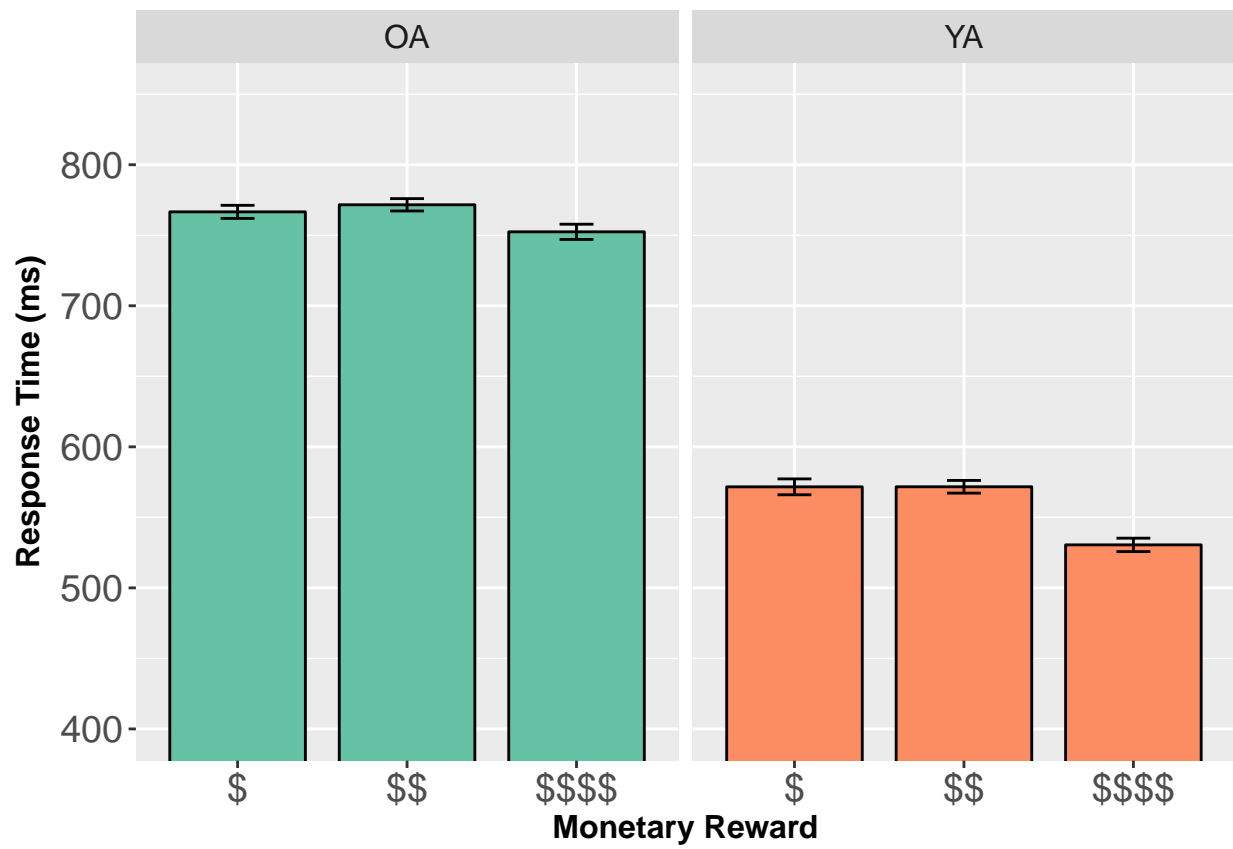
## Plot: RT by age group and money

```
RT.sum=summarySEwithin2(data=rewRT.means, measurevar = "meanRT", withinvars = c("money"),
                        betweenvars = c("group"), idvar = "subID")
pandoc.table(RT.sum)
```

```
##
## -----
##  group    money    N    meanRT    meanRTNormed    sd    se    ci
## -----
##   OA      $      44    766.6      653.3      30.77  4.639  9.355
##
##   OA     $$      44    771.6      658.3      29.02  4.374  8.822
##
##   OA    $$$$      44    752.4      639.1      35.8   5.397  10.88
##
##   YA      $      54    571.6      663.9      41.44  5.639  11.31
##
##   YA     $$      54    571.7      664       32.99  4.49   9.006
##
##   YA    $$$$      54    530.5      622.8      34.94  4.754  9.536
## -----
```

```
p.RT.1<-ggplot(RT.sum, aes(x=money, y=meanRT, fill=group)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
           stat="identity", width=0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
               aes(ymin=meanRT-se, ymax=meanRT+se), width=.2) +
  facet_grid(.~group) +
  xlab("Monetary Reward") + ylab("Response Time (ms)") +
  #ggtitle("Task Performance Across Money") +
  coord_cartesian(ylim=c(400,850)) +
  #scale_fill_discrete(name="Monetary Reward") +
  scale_fill_brewer(palette="Set2") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position="none",
        strip.text.x = element_text(size = 12))
```

p.RT.1



## Generalized Linear Mixed Effects Models of RT by Monetary Reward and Age Group

```
incentiveRT<- incentive %>% filter(ACC==1)

# Monetary Reward x Group Effects
m.RT.money <-lmer(formula = RT ~ moneyCode * groupCode + (1+moneyCode|subID),
  data = incentiveRT, REML = FALSE)
summary(m.RT.money)
```

```
## Linear mixed model fit by maximum likelihood t-tests use Satterthwaite
## approximations to degrees of freedom [lmerMod]
## Formula: RT ~ moneyCode * groupCode + (1 + moneyCode | subID)
## Data: incentiveRT
##
##           AIC          BIC      logLik deviance df.resid
##  318351.4   318415.9 -159167.7   318335.4     23500
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -5.5253 -0.5813 -0.1843  0.3304  6.9152
##
## Random effects:
##   Groups      Name              Variance Std.Dev. Corr
##   subID      (Intercept)  18453.4    135.84
##             moneyCode      427.9     20.69  -0.19
##   Residual                43503.7    208.58
## Number of obs: 23508, groups:  subID, 98
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    557.859    18.583   97.870   30.021 < 2e-16 ***
## moneyCode      -21.130     3.635  100.930   -5.814 7.19e-08 ***
## groupCode      205.569    27.726   97.780    7.414 4.45e-11 ***
## moneyCode:groupCode  13.919     5.380   97.800    2.587 0.0111 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd gropCd
## moneyCode   -0.146
## groupCode   -0.670  0.098
## mnyCd:grpCd  0.098 -0.676 -0.146
```

```
# YA
m.RT.money.YA <-lmer(formula = RT ~ moneyCode + (1+moneyCode|subID),
  data = subset(incentiveRT,group=="YA"))
summary(m.RT.money.YA)
```

```
## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: RT ~ moneyCode + (1 + moneyCode | subID)
## Data: subset(incentiveRT, group == "YA")
##
```

```

## REML criterion at convergence: 165228.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.8377 -0.5686 -0.1826  0.3125  8.2562
##
## Random effects:
##   Groups   Name      Variance Std.Dev. Corr
##   subID    (Intercept) 9260.8   96.23
##           moneyCode    614.1   24.78  -0.30
##   Residual                30669.9 175.13
## Number of obs: 12525, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)  557.869     13.192   52.990  42.289 < 2e-16 ***
## moneyCode   -21.019      3.888   53.990  -5.407 1.5e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr)
## moneyCode -0.262
# OA
m.RT.money.OA <-lmer(formula = RT ~ moneyCode + (1+moneyCode|subID),
                     data = subset(incentive,group=="OA"))
summary(m.RT.money.OA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: RT ~ moneyCode + (1 + moneyCode | subID)
## Data: subset(incentive, group == "OA")
##
## REML criterion at convergence: 180042.8
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4909 -0.5143 -0.1527  0.3869  4.5245
##
## Random effects:
##   Groups   Name      Variance Std.Dev. Corr
##   subID    (Intercept) 16829.1  129.73
##           moneyCode    160.5   12.67   0.11
##   Residual                85479.7 292.37
## Number of obs: 12672, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)  739.524     19.729   43.000  37.484 <2e-16 ***
## moneyCode    -3.792      3.710   43.000  -1.022  0.312
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:

```

```
##          (Intr)
## moneyCode 0.054
```



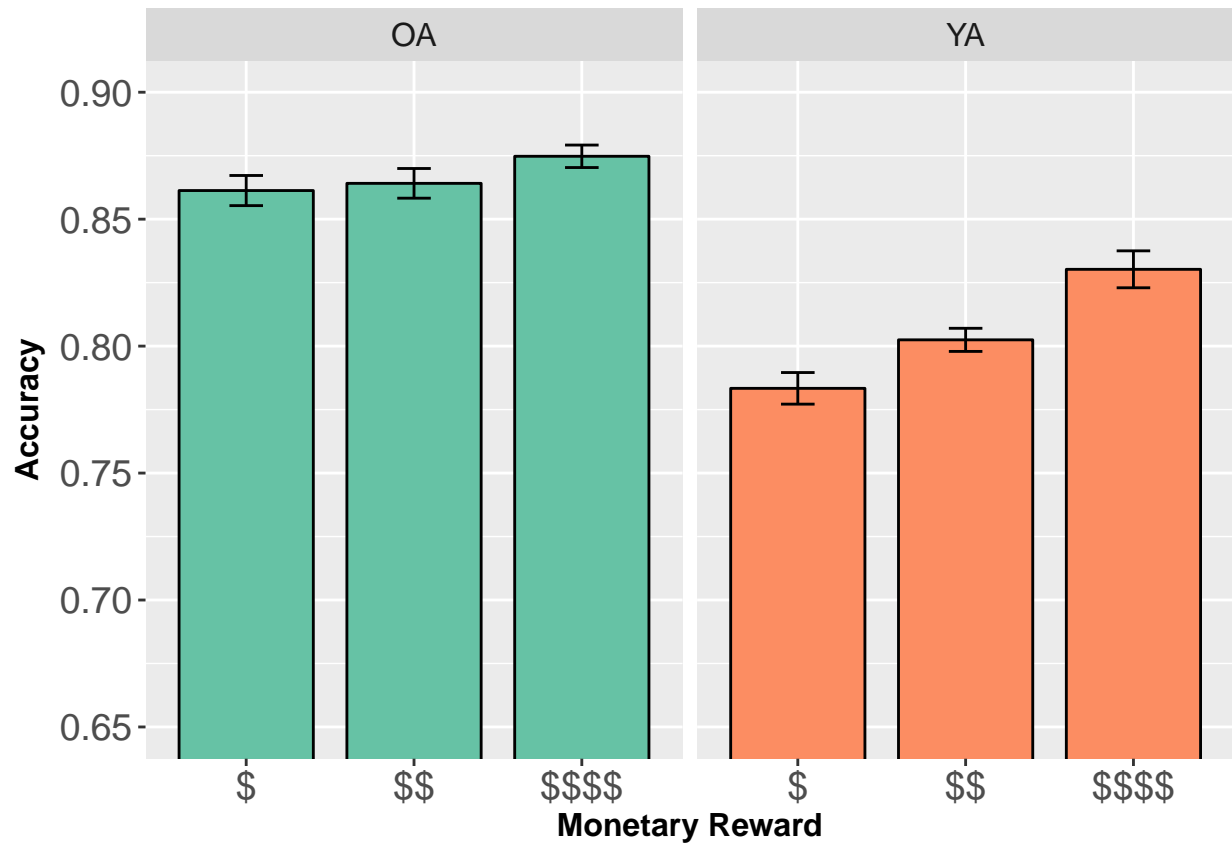
## Plot: Accuracy by age group and money

```
ACC.sum=summarySEwithin2(data=rew.means, measurevar = "meanACC", withinvars = c("money"),
                          betweenvars = c("group"), idvar = "subID")
pandoc.table(ACC.sum)
```

```
##
## -----
##  group  money  N  meanACC  meanACCNormed  sd      se      ci
## -----
##   OA      $   44  0.8613    0.8275      0.03938  0.005936  0.01197
##
##   OA     $$   44  0.8641    0.8303      0.03879  0.005848  0.01179
##
##   OA    $$$$  44  0.8748    0.841       0.02936  0.004426  0.008926
##
##   YA      $   54  0.7834    0.8109      0.04576  0.006227  0.01249
##
##   YA     $$   54  0.8025    0.83        0.03346  0.004554  0.009133
##
##   YA    $$$$  54  0.8302    0.8578      0.05348  0.007277  0.0146
## -----
```

```
p.ACC.1<-ggplot(ACC.sum, aes(x=money, y=meanACC, fill=group)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
           stat="identity", width=0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
               aes(ymin=meanACC-se, ymax=meanACC+se), width=.2) +
  facet_grid(.~group) +
  xlab("Monetary Reward") + ylab("Accuracy") +
  #ggtitle("Task Performance Across Money") +
  coord_cartesian(ylim=c(.65,.9)) +
  #scale_fill_discrete(name="Monetary Reward") +
  scale_fill_brewer(palette="Set2") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position="none",
        strip.text.x = element_text(size = 12))
```

```
p.ACC.1
```



## Generalized Linear Mixed Effects Models of Accuracy by Monetary Reward and Age Group

```
# Monetary Reward
m.ACC.money <-glmer(formula = ACC ~ moneyCode * groupCode + (1+moneyCode|subID),
  data = incentive, family="binomial")
summary(m.ACC.money)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ moneyCode * groupCode + (1 + moneyCode | subID)
## Data: incentive
##
##      AIC      BIC   logLik deviance df.resid
## 23423.8 23481.6 -11704.9 23409.8    28217
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -5.4982  0.2460  0.3464  0.4617  1.5154
##
## Random effects:
## Groups Name      Variance Std.Dev. Corr
## subID (Intercept) 0.54827  0.7405
## moneyCode 0.01096  0.1047  -0.47
## Number of obs: 28224, groups: subID, 98
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.56118    0.10311  15.141 < 2e-16 ***
## moneyCode         0.14528    0.03053   4.759 1.94e-06 ***
## groupCode         0.53770    0.15481   3.473 0.000514 ***
## moneyCode:groupCode -0.10763    0.04830  -2.228 0.025862 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd gropCd
## moneyCode    -0.201
## groupCode    -0.664  0.134
## mnyCd:grpCd  0.128 -0.583 -0.194
```

```
# YA
m.ACC.money.YA <-glmer(formula = ACC ~ moneyCode + (1+moneyCode|subID),
  data = subset(incentive,group=="YA"), family="binomial")
summary(m.ACC.money.YA)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ moneyCode + (1 + moneyCode | subID)
## Data: subset(incentive, group == "YA")
##
##      AIC      BIC   logLik deviance df.resid
```

```

## 14339.3 14377.5 -7164.6 14329.3 15547
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.1840  0.2554  0.4097  0.5006  1.5224
##
## Random effects:
##   Groups Name      Variance Std.Dev. Corr
##   subID  (Intercept) 0.48144  0.6939
##           moneyCode  0.01926  0.1388  -0.45
## Number of obs: 15552, groups:  subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  1.56017    0.09716   16.06 < 2e-16 ***
## moneyCode    0.14068    0.03357    4.19 2.79e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr)
## moneyCode -0.232
# OA
m.ACC.money.OA <-glmer(formula = ACC ~ moneyCode + (1+moneyCode|subID),
                      data = subset(incentive,group=="OA"), family="binomial")
summary(m.ACC.money.OA)

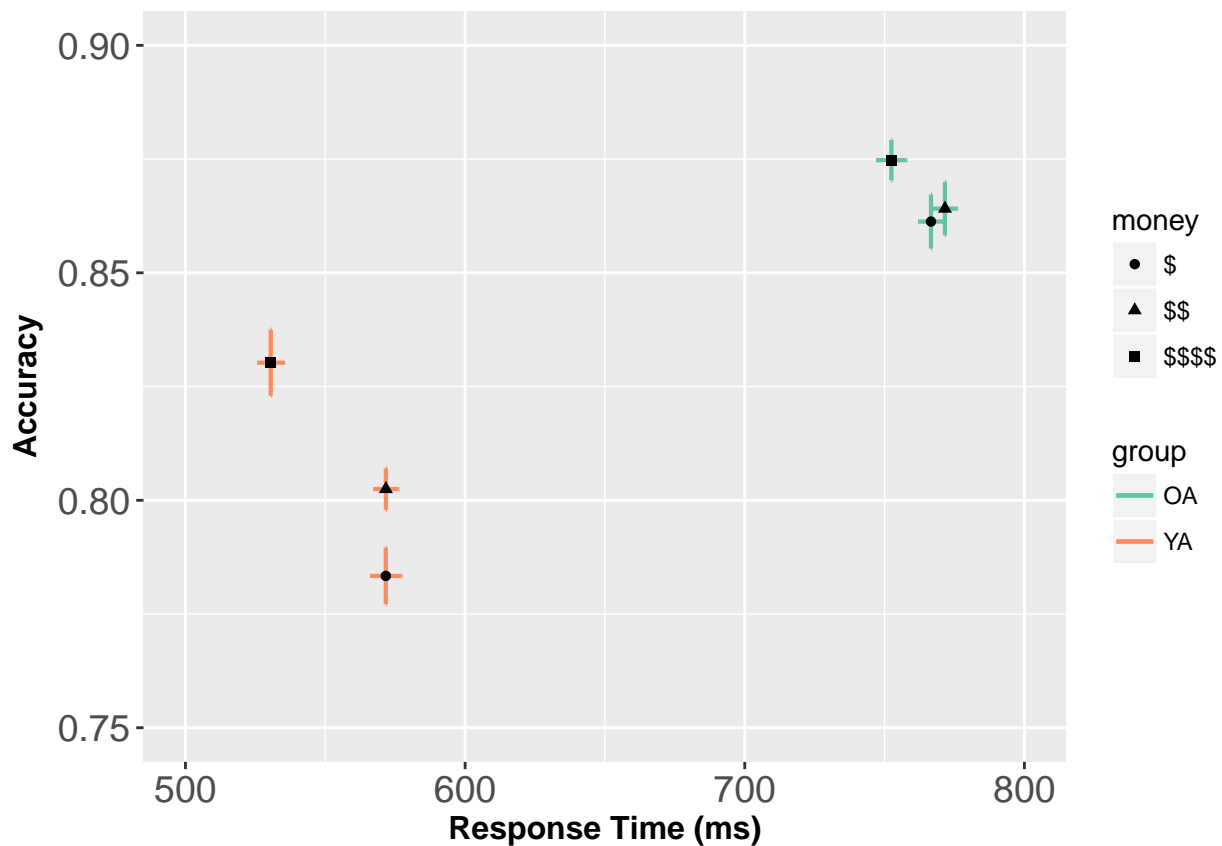
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ moneyCode + (1 + moneyCode | subID)
## Data: subset(incentive, group == "OA")
##
##      AIC      BIC   logLik deviance df.resid
##  9087.0   9124.3  -4538.5   9077.0   12667
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -5.5497  0.2295  0.3177  0.3742  0.9756
##
## Random effects:
##   Groups Name      Variance Std.Dev. Corr
##   subID  (Intercept) 0.632999 0.79561
##           moneyCode  0.001222 0.03495  -1.00
## Number of obs: 12672, groups:  subID, 44
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  2.10104    0.12399   16.945 <2e-16 ***
## moneyCode    0.04530    0.03808    1.189   0.234
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:

```

```
##          (Intr)
## moneyCode -0.128
```

## Combined RT-Accuracy Plot

```
RTACC.sum<-full_join(x = RT.sum, y = ACC.sum, by=c("group","money","N"))
p.RTACC.1<-ggplot(data = RTACC.sum, aes(x = meanRT, y = meanACC, color=group)) +
  geom_errorbar(aes(ymin=meanACC-se.y, ymax=meanACC+se.y), width=0, size=.75) +
  geom_errorbarh(aes(xmin=meanRT-se.x, xmax=meanRT+se.x), height=0, size=.75) +
  geom_point(aes(shape=money), size=1.5, color="black") +
  xlab("Response Time (ms)") + ylab("Accuracy") +
  coord_cartesian(xlim=c(500,800), ylim=c(.75,.9)) +
  scale_color_brewer(palette="Set2") +
  scale_fill_brewer(palette = "Set2") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        #legend.position="none",
        strip.text.x = element_text(size = 12))
p.RTACC.1
```



## Transition Effects From Baseline to Incentive Blocks

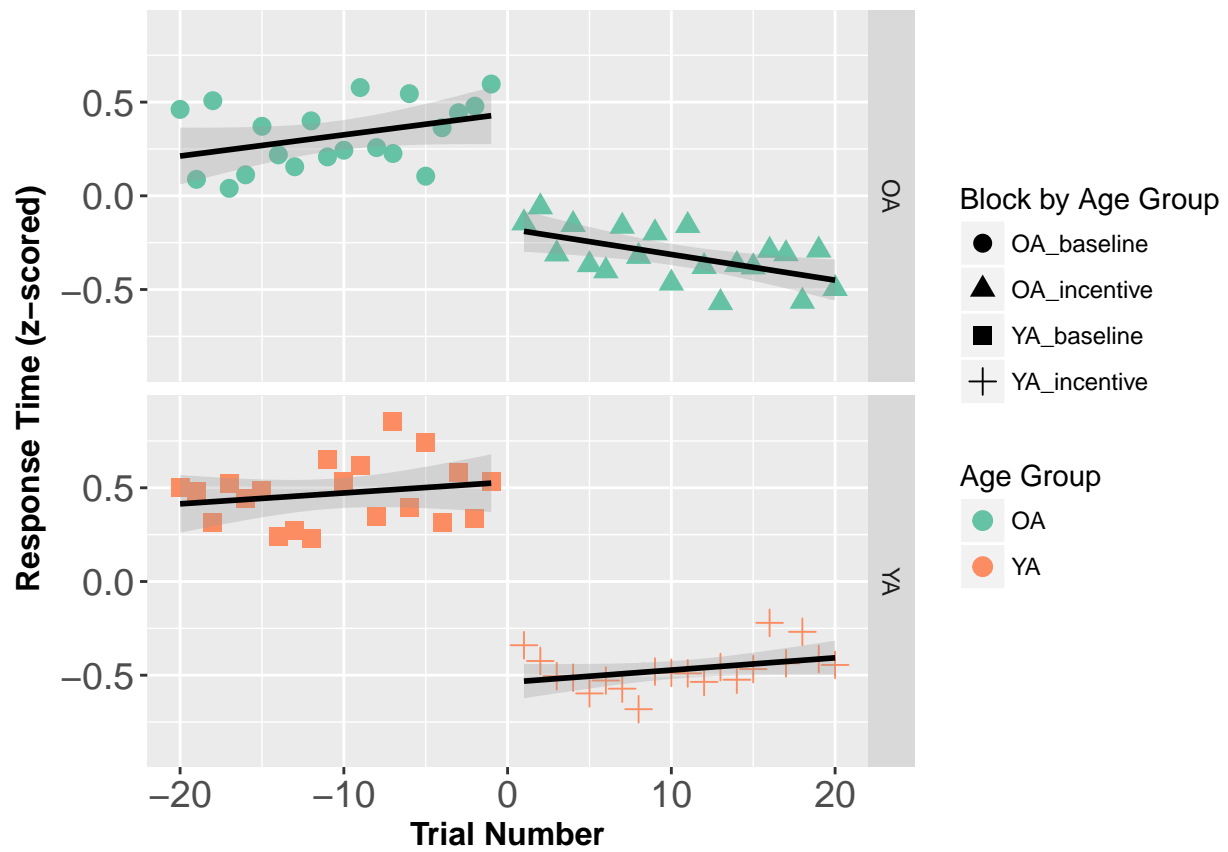
Plot: Transition between last 20 trials of baseline and first 20 trials of first incentive block.

```
transition.base <- data.base %>% select(subID, group, Trial, RT, ACC, condition) %>%
  filter(Trial>=77) %>%
  mutate(Trial=Trial-97)
transition.reward <- incentive %>% filter(block==3 & Trial<=20) %>%
  select(subID, group, Trial, RT, ACC) %>%
  mutate(condition="incentive")
transition=rbind(transition.base,transition.reward)
transition <- transition %>% arrange(subID, condition, Trial) %>%
  group_by(subID) %>%
  mutate(RT_zscore = scale(RT),
         ACC=as.numeric(ACC),
         groupCode=factor(group, levels=c("YA","OA"), labels=c(0,1)),
         conditionCode=factor(condition, levels=c("baseline","incentive"), labels=c(1,0)))
transition$groupCode<-as.numeric(levels(transition$groupCode)[transition$groupCode])
transition$conditionCode<-as.numeric(levels(transition$conditionCode)[transition$conditionCode])
ix_NA<-which(transition$ACC==0)
transition$RT[ix_NA]<-""

trans.sum<-summarySEwithin2(data = transition, measurevar = "RT_zscore", betweenvars = "group",
                           withinvars = c("Trial","condition"), idvar = "subID", na.rm = TRUE)

## Automatically converting the following non-factors to factors: Trial
trans.sum <- trans.sum %>% arrange(group,Trial) %>%
  unite("group_cond", c("group","condition"), sep="_", remove = FALSE)
trans.sum$Trial<-as.numeric(levels(trans.sum$Trial)[trans.sum$Trial])

# Plotting z-scored RT
p.trans.1<-ggplot(data = trans.sum, mapping = aes(x = Trial, y = RT_zscore, shape=group_cond)) +
  geom_point(aes(color=group),size=3) +
  geom_smooth(method="lm", aes(group=group_cond), color="black", alpha=.3) +
  facet_grid(group~.) +
  coord_cartesian(ylim=c(-.9,.9)) +
  scale_color_brewer(palette="Set2") +
  scale_fill_brewer(palette = "Set2") +
  xlab("Trial Number") + ylab("Response Time (z-scored)") +
  labs(shape = "Block by Age Group", color = "Age Group") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face="bold"),
        axis.text=element_text(size=14),
        #legend.position="none",
        #legend.title=element_text(),
        strip.text.x = element_text(size = 10))
p.trans.1
```





## GLM: RT by trial, condition, and age group

Transition between last 20 trials of baseline and first 20 trials of first incentive block.

```
# GLM on zscored RT
m.trans <-lmer(formula = RT_zscore ~ Trial * conditionCode * groupCode + (1|subID),
              data = transition)
summary(m.trans)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
##   to degrees of freedom [lmerMod]
## Formula: RT_zscore ~ Trial * conditionCode * groupCode + (1 | subID)
##   Data: transition
##
## REML criterion at convergence: 10327
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.3958 -0.5923 -0.1487  0.4729  5.2296
##
## Random effects:
##   Groups      Name              Variance Std.Dev.
##   subID      (Intercept) 1.157e-32 1.076e-16
##   Residual                8.060e-01 8.978e-01
## Number of obs: 3920, groups: subID, 99
##
## Fixed effects:
##
##              Estimate Std. Error    df t value
## (Intercept)   -5.382e-01  5.675e-02 3.912e+03  -9.483
## Trial           6.542e-03  4.738e-03 3.912e+03   1.381
## conditionCode  1.069e+00  8.026e-02 3.912e+03  13.315
## groupCode      3.625e-01  8.470e-02 3.912e+03   4.280
## Trial:conditionCode -7.316e-04  6.700e-03 3.912e+03  -0.109
## Trial:groupCode   -2.026e-02  7.070e-03 3.912e+03  -2.866
## conditionCode:groupCode -4.542e-01  1.198e-01 3.912e+03  -3.792
## Trial:conditionCode:groupCode 2.578e-02  9.999e-03 3.912e+03   2.578
##
##              Pr(>|t|)
## (Intercept)    < 2e-16 ***
## Trial           0.167391
## conditionCode    < 2e-16 ***
## groupCode       1.92e-05 ***
## Trial:conditionCode 0.913049
## Trial:groupCode   0.004184 **
## conditionCode:groupCode 0.000151 ***
## Trial:conditionCode:groupCode 0.009963 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Trial  cndtnC  grpCd  Trl:cC  Trl:gC  cndC:C
## Trial          -0.877
## conditionCd  -0.707  0.620
## groupCode    -0.670  0.587  0.474
## Trl:cndtnCd  0.620 -0.707  0.000 -0.415
## Trial:grpCd   0.587 -0.670 -0.415 -0.877  0.474
```

```
## cndtnCd:grC  0.474 -0.415 -0.670 -0.707  0.000  0.620
## Tr1:cndtC:C -0.415  0.474  0.000  0.620 -0.670 -0.707  0.000

# YA only
m.trans.YA <-lmer(formula = RT_zscore ~ Trial * conditionCode + (1|subID),
                  data = subset(transition, group=="YA"))
summary(m.trans.YA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: RT_zscore ~ Trial * conditionCode + (1 | subID)
## Data: subset(transition, group == "YA")
##
## REML criterion at convergence: 5546.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4800 -0.5912 -0.1485  0.4322  5.4043
##
## Random effects:
##   Groups   Name                Variance Std.Dev.
##   subID    (Intercept) 1.037e-32 1.018e-16
##   Residual                        7.547e-01 8.687e-01
## Number of obs: 2160, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   -5.382e-01  5.492e-02 2.156e+03  -9.800   <2e-16 ***
## Trial           6.542e-03  4.584e-03 2.156e+03   1.427    0.154
## conditionCode  1.069e+00  7.766e-02 2.156e+03  13.760   <2e-16 ***
## Trial:conditionCode -7.316e-04  6.483e-03 2.156e+03  -0.113    0.910
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Trial  cndtnC
## Trial          -0.877
## conditionCd -0.707  0.620
## Tr1:cndtnCd  0.620 -0.707  0.000

# OA only
m.trans.OA <-lmer(formula = RT_zscore ~ Trial * conditionCode + (1|subID),
                  data = subset(transition, group=="OA"))
summary(m.trans.OA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: RT_zscore ~ Trial * conditionCode + (1 | subID)
## Data: subset(transition, group == "OA")
##
## REML criterion at convergence: 4770.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.2336 -0.5949 -0.1491  0.5297  4.7458
##
```

```

## Random effects:
##   Groups   Name      Variance Std.Dev.
## subID    (Intercept) 1.310e-32 1.144e-16
## Residual                8.689e-01 9.321e-01
## Number of obs: 1760, groups: subID, 45
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   -1.757e-01  6.528e-02 1.756e+03  -2.692  0.00718 **
## Trial          -1.372e-02  5.449e-03 1.756e+03  -2.518  0.01191 *
## conditionCode    6.144e-01  9.232e-02 1.756e+03   6.656 3.76e-11 ***
## Trial:conditionCode 2.505e-02  7.707e-03 1.756e+03   3.250  0.00117 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Trial  cndtnC
## Trial          -0.877
## conditionCd -0.707  0.620
## Trl:cndtnCd  0.620 -0.707  0.000

# OA baseline
m.trans.OA.base <-lmer(formula = RT_zscore ~ Trial * conditionCode + (1|subID),
                      data = subset(transition, group=="OA" & condition=="baseline"))

## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
summary(m.trans.OA.base)

## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: RT_zscore ~ Trial * conditionCode + (1 | subID)
## Data: subset(transition, group == "OA" & condition == "baseline")
##
## REML criterion at convergence: 2629.3
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.6844 -0.6174 -0.1337  0.5681  4.1301
##
## Random effects:
##   Groups   Name      Variance Std.Dev.
## subID    (Intercept) 6.692e-21 8.180e-11
## Residual                1.147e+00 1.071e+00
## Number of obs: 880, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)  4.387e-01  7.501e-02 8.780e+02   5.849 6.98e-09 ***
## Trial         1.133e-02  6.262e-03 8.780e+02   1.809  0.0707 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
##
## Correlation of Fixed Effects:
##      (Intr)
## Trial 0.877
## fit warnings:
## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
# OA incentive
m.trans.OA.incentive <-lmer(formula = RT_zscore ~ Trial * conditionCode + (1|subID),
  data = subset(transition, group=="OA" & condition=="incentive"))

## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
summary(m.trans.OA.incentive)

## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: RT_zscore ~ Trial * conditionCode + (1 | subID)
## Data: subset(transition, group == "OA" & condition == "incentive")
##
## REML criterion at convergence: 2042.9
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.4097 -0.6102 -0.1643  0.5044  5.5568
##
## Random effects:
## Groups   Name            Variance Std.Dev.
## subID    (Intercept) 0.01358  0.1165
## Residual                0.57726  0.7598
## Number of obs: 880, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept) -0.175706   0.056032 361.100000  -3.136  0.00185 **
## Trial        -0.013719   0.004442 835.000000  -3.089  0.00208 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr)
## Trial -0.832
## fit warnings:
## fixed-effect model matrix is rank deficient so dropping 2 columns / coefficients
```

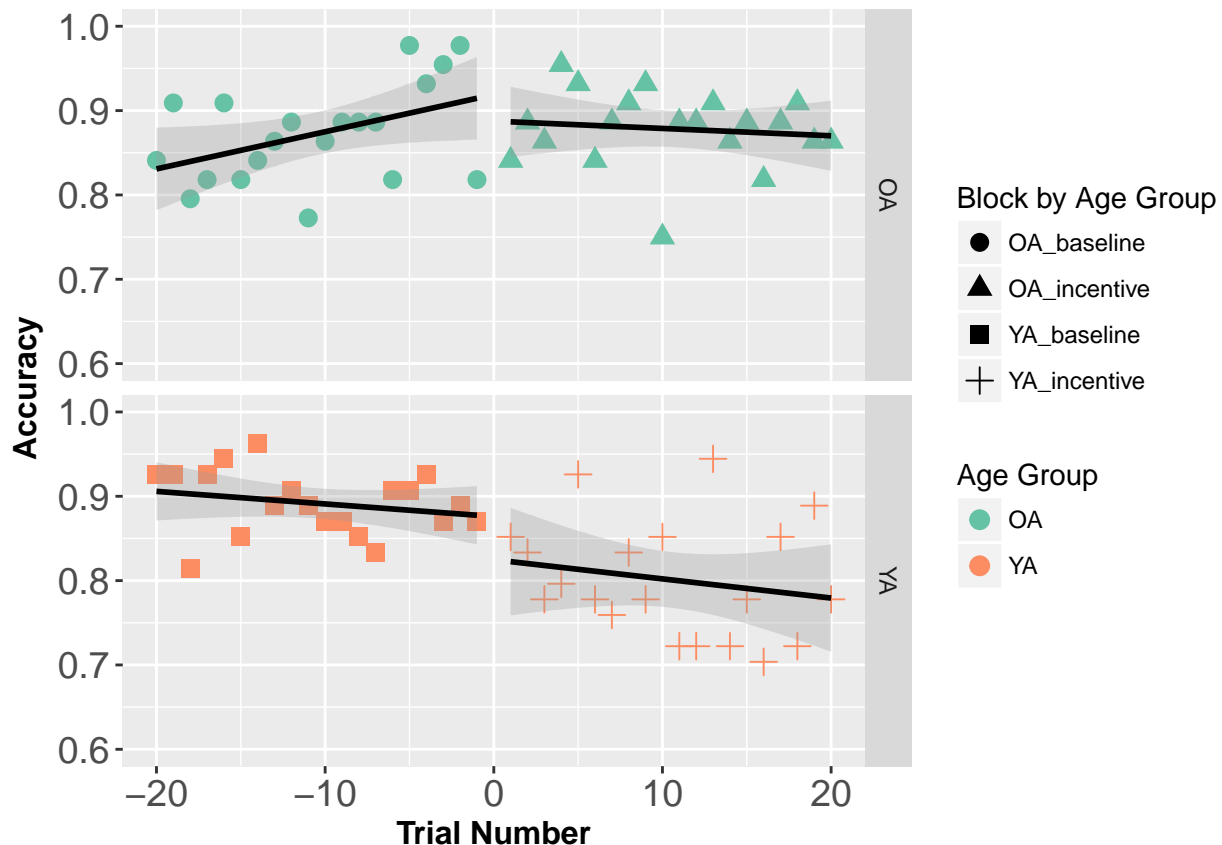
## Plot: Transition of Accuracy between baseline and incentive block

```
trans.sum<-summarySEwithin2(data = transition, measurevar = "ACC", betweenvars = "group",
  withinvars = c("Trial","condition"), idvar = "subID", na.rm = TRUE)

## Automatically converting the following non-factors to factors: Trial
```

```
trans.sum <- trans.sum %>% arrange(group,Trial) %>%
  unite("group_cond", c("group","condition"), sep="_", remove = FALSE)
trans.sum$Trial<-as.numeric(levels(trans.sum$Trial)[trans.sum$Trial])

# Plotting Accuracy
p.trans.2<-ggplot(data = trans.sum, mapping = aes(x = Trial, y = ACC, shape=group_cond)) +
  geom_point(aes(color=group),size=3) +
  geom_smooth(method="lm", aes(group=group_cond), color="black", alpha=.3) +
  facet_grid(group~.) +
  coord_cartesian(ylim=c(0.6,1)) +
  scale_color_brewer(palette="Set2") +
  scale_fill_brewer(palette = "Set2") +
  xlab("Trial Number") + ylab("Accuracy") +
  labs(shape = "Block by Age Group", color = "Age Group") +
  theme(plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face="bold"),
        axis.text=element_text(size=14),
        #legend.position="none",
        #legend.title=element_text(),
        strip.text.x = element_text(size = 10))
p.trans.2
```



## GLM: Accuracy by trial, condition, and age group

Transition between last 20 trials of baseline and first 20 trials of first incentive block.

```

# GLM on zscored ACC
m.trans.ACC <-glmer(formula = ACC ~ Trial * conditionCode * groupCode + (1|subID),
  data = transition, family="binomial")

## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control
## $checkConv, : Model failed to converge with max|grad| = 0.0132619 (tol =
## 0.001, component 1)

summary(m.trans.ACC)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ Trial * conditionCode * groupCode + (1 | subID)
## Data: transition
##
##      AIC      BIC   logLik deviance df.resid
## 2957.4   3013.8 -1469.7  2939.4     3911
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.8049  0.2254  0.3063  0.3948  1.2049
##
## Random effects:
## Groups Name      Variance Std.Dev.
## subID (Intercept) 0.7367   0.8583
## Number of obs: 3920, groups: subID, 99
##
## Fixed effects:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.7472164  0.2069024   8.445  <2e-16 ***
## Trial             -0.0158345  0.0137817  -1.149   0.2506
## conditionCode     0.4423966  0.2632333   1.681   0.0928 .
## groupCode         0.6429966  0.3343616   1.923   0.0545 .
## Trial:conditionCode -0.0007676  0.0221812  -0.035   0.9724
## Trial:groupCode     0.0066958  0.0231625   0.289   0.7725
## conditionCode:groupCode -0.1049462  0.4200173  -0.250   0.8027
## Trial:conditionCode:groupCode 0.0548964  0.0343923   1.596   0.1104
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Trial  cndtnC  gropCd  Trl:cC  Trl:gC  cndC:C
## Trial          -0.719
## conditionCd   -0.517  0.563
## groupCode     -0.612  0.444  0.321
## Trl:cndtnCd   0.446 -0.621  0.174 -0.276
## Trial:grpCd    0.427 -0.595 -0.335 -0.743  0.370
## cndtnCd:grC   0.324 -0.353 -0.627 -0.564 -0.109  0.591
## Trl:cndtC:C  -0.286  0.400 -0.112  0.503 -0.645 -0.674  0.085
## convergence code: 0
## Model failed to converge with max|grad| = 0.0132619 (tol = 0.001, component 1)

# YA only
m.trans.ACC.YA <-glmer(formula = ACC ~ Trial * conditionCode + (1|subID),

```

```
data = subset(transition, group=="YA"), family="binomial")
summary(m.trans.ACC.YA)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ Trial * conditionCode + (1 | subID)
## Data: subset(transition, group == "YA")
##
##      AIC      BIC    logLik deviance df.resid
##  1739.4   1767.8   -864.7   1729.4     2155
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.6821  0.2358  0.3289  0.4332  1.1777
##
## Random effects:
## Groups Name      Variance Std.Dev.
## subID (Intercept) 0.6118   0.7822
## Number of obs: 2160, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    1.7307084  0.2013711   8.595  <2e-16 ***
## Trial           -0.0157202  0.0137502  -1.143   0.2529
## conditionCode    0.4398910  0.2628023   1.674   0.0942 .
## Trial:conditionCode -0.0007862  0.0221448  -0.036   0.9717
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) Trial  cndtnC
## Trial          -0.737
## conditionCd  -0.529  0.562
## Trl:cndtnCd   0.457 -0.621  0.175
```

```
# OA only
m.trans.ACC.OA <-glmer(formula = ACC ~ Trial * conditionCode + (1|subID),
data = subset(transition, group=="OA"), family="binomial")
summary(m.trans.ACC.OA)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ Trial * conditionCode + (1 | subID)
## Data: subset(transition, group == "OA")
##
##      AIC      BIC    logLik deviance df.resid
##  1219.0   1246.4   -604.5   1209.0     1755
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.7496  0.2154  0.2840  0.3613  1.1949
##
```

```

## Random effects:
##   Groups Name      Variance Std.Dev.
##   subID (Intercept) 0.9054   0.9515
## Number of obs: 1760, groups:  subID, 45
##
## Fixed effects:
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      2.420810   0.273901   8.838   <2e-16 ***
## Trial             -0.009221   0.018659  -0.494   0.6212
## conditionCode     0.340209   0.328025   1.037   0.2997
## Trial:conditionCode 0.054605   0.026349   2.072   0.0382 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr) Trial  cndtnC
## Trial      -0.730
## conditionCd -0.568  0.608
## Trl:cndtnCd 0.524 -0.709  0.025

# OA only baseline
m.trans.ACC.OA.baseline <-glmer(formula = ACC ~ Trial + (1|subID),
                                data = subset(transition, group=="OA" & condition=="baseline"), family="binomial")
summary(m.trans.ACC.OA.baseline)

## Generalized linear mixed model fit by maximum likelihood (Laplace
##   Approximation) [glmerMod]
##   Family: binomial ( logit )
## Formula: ACC ~ Trial + (1 | subID)
##   Data: subset(transition, group == "OA" & condition == "baseline")
##
##      AIC      BIC   logLik deviance df.resid
##   584.3    598.6   -289.1    578.3     877
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.8457  0.1651  0.2281  0.3211  1.4316
##
## Random effects:
##   Groups Name      Variance Std.Dev.
##   subID (Intercept) 1.809    1.345
## Number of obs: 880, groups:  subID, 44
##
## Fixed effects:
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      3.08108   0.35539   8.670   <2e-16 ***
## Trial             0.04998   0.01997   2.503   0.0123 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr)
## Trial 0.668

```



```

# OA only incentive
m.trans.ACC.OA.incentive <-glmer(formula = ACC ~ Trial + (1|subID),
    data = subset(transition, group=="OA" & condition=="incentive"), family="binomial")

## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control
## $checkConv, : Model failed to converge with max|grad| = 0.0488536 (tol =
## 0.001, component 1)

summary(m.trans.ACC.OA.incentive)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: ACC ~ Trial + (1 | subID)
## Data: subset(transition, group == "OA" & condition == "incentive")
##
##      AIC      BIC   logLik deviance df.resid
##  639.2    653.5   -316.6    633.2      877
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.5938  0.2543  0.2997  0.3688  0.7749
##
## Random effects:
## Groups Name          Variance Std.Dev.
## subID (Intercept) 0.5531    0.7437
## Number of obs: 880, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  2.276739   0.001444   1576 < 2e-16 ***
## Trial        -0.008720   0.001445     -6 1.59e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr)
## Trial 0.000
## convergence code: 0
## Model failed to converge with max|grad| = 0.0488536 (tol = 0.001, component 1)

```

# Integration of Primary and Secondary Motivational Incentives in Older and Younger Adults

Plot: Reward rate by monetary reward, liquid, and age group

```
RR.sum=summarySEwithin2(data=incentive9.means, measurevar = "meanRR",
                        withinvars = c("money","liquid"), betweenvars = c("group"),
                        idvar = "subID")
pandoc.table(RR.sum)
```

```
##
## -----
##  group    money    liquid    N    meanRR    meanRRNormed    sd    se    ci
## -----
##  OA      $      juice    44    0.5788    0.6161    0.08623    0.013    0.02622
##
##  OA      $      neutral  44    0.5611    0.5984    0.1102    0.01662    0.03351
##
##  OA      $      saltwater 44    0.5582    0.5955    0.1138    0.01715    0.03459
##
##  OA     $$      juice    44    0.5682    0.6055    0.1097    0.01654    0.03335
##
##  OA     $$      neutral  44    0.5554    0.5927    0.1206    0.01818    0.03667
##
##  OA     $$      saltwater 44    0.5561    0.5934    0.1135    0.01711    0.03451
##
##  OA     $$$$     juice    44    0.6108    0.6481    0.127    0.01915    0.03861
##
##  OA     $$$$     neutral  44    0.6009    0.6381    0.104    0.01567    0.0316
##
##  OA     $$$$     saltwater 44    0.6023    0.6396    0.1281    0.01931    0.03895
##
##  YA      $      juice    54    0.6696    0.6392    0.09627    0.0131    0.02628
##
##  YA      $      neutral  54    0.6256    0.5952    0.1288    0.01753    0.03516
##
##  YA      $      saltwater 54    0.5168    0.4864    0.1689    0.02299    0.0461
##
##  YA     $$      juice    54    0.6678    0.6374    0.1336    0.01818    0.03645
##
##  YA     $$      neutral  54    0.6256    0.5952    0.1175    0.016    0.03208
##
##  YA     $$      saltwater 54    0.5602    0.5298    0.1301    0.0177    0.0355
##
##  YA     $$$$     juice    54    0.7523    0.7219    0.09888    0.01346    0.02699
##
##  YA     $$$$     neutral  54    0.7216    0.6913    0.09788    0.01332    0.02672
##
##  YA     $$$$     saltwater 54    0.6615    0.6311    0.1081    0.01471    0.0295
## -----
```

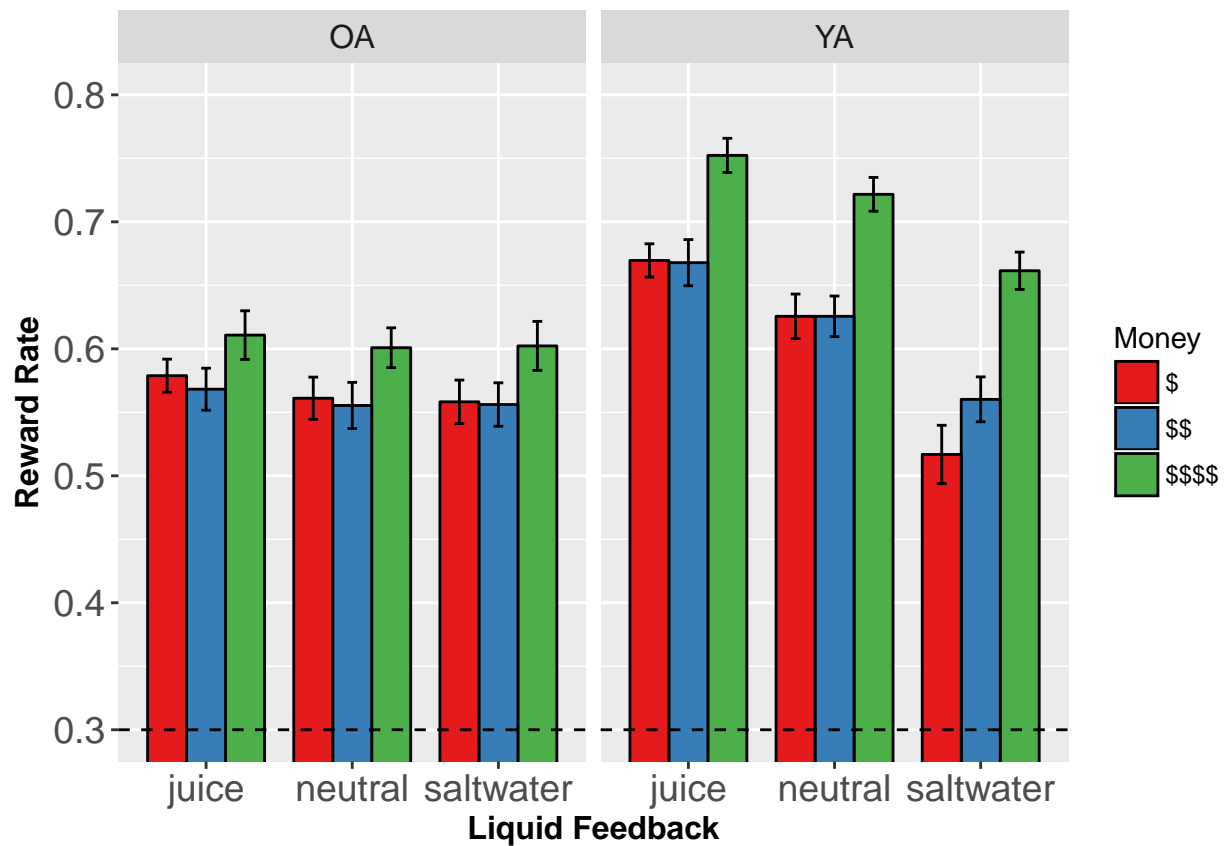
```
# first plot
p.RR.2<-ggplot(RR.sum, aes(x=liquid, y=meanRR, fill=money)) +
```

```

geom_bar(position=position_dodge(width=0.8), color="black",
         stat="identity", width=0.8) +
geom_errorbar(position=position_dodge(width=0.8),
             aes(ymin=meanRR-se, ymax=meanRR+se), width=.2) +
xlab("Liquid Feedback") + ylab("Reward Rate") +
labs(fill = "Money") +
facet_grid(.~group) +
geom_hline(yintercept = .30, linetype="dashed") +
scale_fill_brewer(palette="Set1") +
coord_cartesian(ylim=c(.3,.8)) +
#scale_fill_discrete(name="Monetary Reward") +
theme(#plot.title=element_text(size=22,face="bold", vjust=2),
      axis.title=element_text(size=12,face = "bold"),
      axis.text=element_text(size=14),
      legend.position="right",
      strip.text.x = element_text(size = 12))

```

p.RR.2



## Analysis: Reward Rate by money, liquid, and age group

```
# Monetary Reward
m.RR.mot9<-glmer(formula = subRewarded ~ moneyCode * liqCode * groupCode +
                  (1+moneyCode+liqCode|subID),
                  data = incentive, family="binomial")
summary(m.RR.mot9)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode * liqCode * groupCode + (1 + moneyCode +
## liqCode | subID)
## Data: incentive
##
##          AIC          BIC    logLik deviance df.resid
## 35732.2 35847.7 -17852.1 35704.2    28210
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.8151 -1.0627  0.5715  0.7703  5.3667
##
## Random effects:
## Groups Name          Variance Std.Dev. Corr
## subID (Intercept) 0.26075  0.5106
## moneyCode 0.01387  0.1178  -0.20
## liqCode 0.11329  0.3366  -0.24  0.53
## Number of obs: 28224, groups: subID, 98
##
## Fixed effects:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.63846   0.07178   8.894 < 2e-16 ***
## moneyCode         0.25760   0.02702   9.534 < 2e-16 ***
## liqCode           0.28281   0.05083   5.564 2.64e-08 ***
## groupCode        -0.31271   0.10692  -2.925 0.003448 **
## moneyCode:liqCode -0.06045   0.02669  -2.265 0.023530 *
## moneyCode:groupCode -0.17426   0.03955  -4.407 1.05e-05 ***
## liqCode:groupCode -0.25305   0.07538  -3.357 0.000788 ***
## moneyCode:liqCode:groupCode 0.04840   0.03866   1.252 0.210587
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd liqCod gropCd mnyCd:lC mnyCd:gC lqCd:C
## moneyCode    -0.105
## liqCode       -0.208  0.292
## groupCode     -0.672  0.071  0.140
## monyCd:lqCd   0.005  0.032  0.021 -0.003
## mnyCd:grpCd   0.072 -0.683 -0.200 -0.113 -0.021
## liqCd:grpCd   0.141 -0.197 -0.674 -0.213 -0.014  0.297
## mnyCd:lqC:C  -0.003 -0.021 -0.014  0.003 -0.690  0.014  0.011

# YA
m.RR.mot9.YA <-glmer(formula = subRewarded ~ moneyCode * liqCode +
```

```

                                (1+moneyCode+liqCode|subID),
                                data = subset(incentive,group=="YA"), family="binomial")
summary(m.RR.mot9.YA)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode * liqCode + (1 + moneyCode + liqCode |
## subID)
## Data: subset(incentive, group == "YA")
##
##      AIC      BIC   logLik deviance df.resid
## 19053.6 19130.1 -9516.8 19033.6    15542
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.917 -1.060  0.540  0.723  5.970
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.29499  0.5431
## moneyCode 0.02097  0.1448  -0.13
## liqCode 0.15128  0.3890  -0.37  0.59
## Number of obs: 15552, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.63853    0.07606   8.396 < 2e-16 ***
## moneyCode      0.26060    0.02942   8.857 < 2e-16 ***
## liqCode        0.28410    0.05738   4.951 7.37e-07 ***
## moneyCode:liqCode -0.05965    0.02677  -2.228  0.0259 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd liqCod
## moneyCode   -0.073
## liqCode     -0.327  0.373
## monyCd:liqCd  0.005  0.030  0.019

# Juice vs Neutral
m.RR.mot9.YA.JvN <-glmer(formula = subRewarded ~ moneyCode * liqCode +
                        (1+moneyCode+liqCode|subID),
                        data = subset(incentive,group=="YA" && liquid!="saltwater"),
                        family="binomial")
summary(m.RR.mot9.YA.JvN)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode * liqCode + (1 + moneyCode + liqCode |
## subID)
## Data: subset(incentive, group == "YA" && liquid != "saltwater")
##

```

```

##      AIC      BIC   logLik deviance df.resid
## 35759.3 35841.8 -17869.6 35739.3   28214
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.9996 -1.0548  0.5694  0.7729  5.1123
##
## Random effects:
##   Groups Name      Variance Std.Dev. Corr
##   subID  (Intercept) 0.28475  0.5336
##           moneyCode  0.02149  0.1466   0.03
##           liqCode    0.12923  0.3595  -0.11  0.61
## Number of obs: 28224, groups:  subID, 98
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.49829     0.05544   8.988 < 2e-16 ***
## moneyCode         0.17942     0.02164   8.292 < 2e-16 ***
## liqCode           0.16957     0.03965   4.276 1.9e-05 ***
## moneyCode:liqCode -0.03710     0.01932  -1.920  0.0548 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd liqCod
## moneyCode      0.024
## liqCode        -0.096  0.388
## monyCd:liqCd   0.003  0.017  0.012

# Saltwater vs Neutral
m.RR.mot9.YA.SvN <-glmer(formula = subRewarded ~ moneyCode * liqCode +
                        (1+moneyCode+liqCode|subID),
                        data = subset(incentive,group=="YA" & liquid!="juice"),
                        family="binomial")
summary(m.RR.mot9.YA.SvN)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode * liqCode + (1 + moneyCode + liqCode |
##          subID)
## Data: subset(incentive, group == "YA" & liquid != "juice")
##
##      AIC      BIC   logLik deviance df.resid
## 12849.7 12922.2 -6414.9 12829.7   10358
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.5054 -1.0367  0.5555  0.7256  6.6775
##
## Random effects:
##   Groups Name      Variance Std.Dev. Corr
##   subID  (Intercept) 0.35580  0.5965
##           moneyCode  0.05532  0.2352  -0.36
##           liqCode    0.35280  0.5940   0.06  0.17

```

```

## Number of obs: 10368, groups:  subID, 54
##
## Fixed effects:
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.69132    0.08689   7.956 1.77e-15 ***
## moneyCode      0.23193    0.04957   4.679 2.89e-06 ***
## liqCode        0.36250    0.09188   3.945 7.96e-05 ***
## moneyCode:liqCode -0.11141    0.05315  -2.096  0.0361 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd liqCod
## moneyCode   -0.205
## liqCode      0.169  0.107
## monyCd:liqCd  0.012  0.541  0.015
# OA
m.RR.mot9.OA <-glmer(formula = subRewarded ~ moneyCode * liqCode +
                     (1+moneyCode+liqCode|subID),
                     data = subset(incentive,group=="OA"), family="binomial")
summary(m.RR.mot9.OA)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ moneyCode * liqCode + (1 + moneyCode + liqCode |
## subID)
## Data: subset(incentive, group == "OA")
##
##      AIC      BIC    logLik deviance df.resid
## 16681.9 16756.4 -8330.9 16661.9    12662
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.1172 -1.0652  0.6329  0.8276  2.1987
##
## Random effects:
## Groups Name          Variance Std.Dev. Corr
## subID  (Intercept)  0.223904  0.47319
##          moneyCode   0.006182  0.07862  -0.39
##          liqCode     0.073080  0.27033  -0.03  0.37
## Number of obs: 12672, groups:  subID, 44
##
## Fixed effects:
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.32497    0.07373   4.408 1.05e-05 ***
## moneyCode      0.08277    0.02567   3.225  0.00126 **
## liqCode        0.03034    0.04673   0.649  0.51620
## moneyCode:liqCode -0.01235    0.02793  -0.442  0.65841
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) monyCd liqCod

```

```
## moneyCode    -0.173
## liqCode      -0.021  0.149
## monyCd:liqCd  0.000  0.003  0.002
```

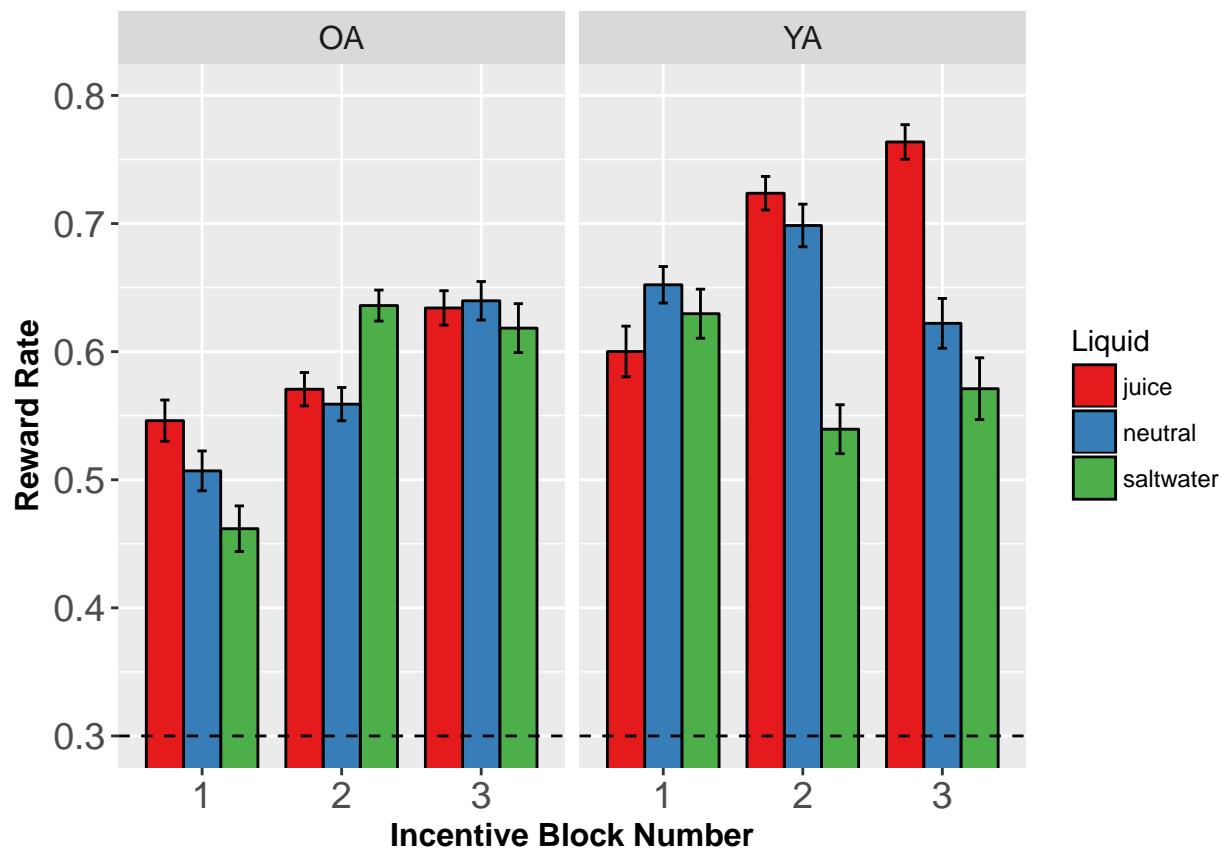
## Plot: mean reward rate by block, money, and age group

```
RR.sum=summarySEwithin2(data=incentive9.means, measurevar = "meanRR",
                        withinvars = c("block","liquid"), betweenvars = c("group"),
                        idvar = "subID")
pandoc.table(RR.sum)
```

```
##
## -----
##  group    block    liquid    N    meanRR    meanRRNormed    sd    se    ci
## -----
##  OA        1      juice     42    0.5461     0.5497     0.1044  0.01611  0.03253
##
##  OA        1      neutral   45    0.5069     0.5378     0.1042  0.01553  0.0313
##
##  OA        1      saltwater 45    0.4618     0.5371     0.1192  0.01776  0.0358
##
##  OA        2      juice     42    0.5707     0.6323     0.08449  0.01304  0.02633
##
##  OA        2      neutral   36    0.559      0.6347     0.07792  0.01299  0.02637
##
##  OA        2      saltwater 54    0.636      0.6288     0.0889   0.0121   0.02427
##
##  OA        3      juice     48    0.6341     0.6797     0.0932   0.01345  0.02706
##
##  OA        3      neutral   51    0.6397     0.6556     0.1073   0.01502  0.03017
##
##  OA        3      saltwater 33    0.6184     0.6767     0.1096   0.01908  0.03886
##
##  YA        1      juice     54    0.6001     0.6344     0.1453   0.01977  0.03966
##
##  YA        1      neutral   54    0.6522     0.6082     0.1047   0.01424  0.02857
##
##  YA        1      saltwater 54    0.6296     0.5482     0.1408   0.01917  0.03844
##
##  YA        2      juice     51    0.7237     0.6724     0.09358  0.0131   0.02632
##
##  YA        2      neutral   54    0.6985     0.6444     0.1224   0.01665  0.0334
##
##  YA        2      saltwater 57    0.5395     0.5502     0.144    0.01907  0.03821
##
##  YA        3      juice     57    0.7637     0.6907     0.1017   0.01347  0.02698
##
##  YA        3      neutral   54    0.6221     0.629      0.1427   0.01942  0.03894
##
##  YA        3      saltwater 51    0.5711     0.5488     0.1723   0.02413  0.04847
## -----
```



```
# first plot
p.RR.3<-ggplot(RR.sum, aes(x=block, y=meanRR, fill=liquid)) +
  geom_bar(position=position_dodge(width=0.8), color="black",
    stat="identity", width=0.8) +
  geom_errorbar(position=position_dodge(width=0.8),
    aes(ymin=meanRR-se, ymax=meanRR+se), width=.2) +
  xlab("Incentive Block Number") + ylab("Reward Rate") +
  labs(fill = "Liquid") +
  facet_grid(.~group) +
  geom_hline(yintercept = .30, linetype="dashed") +
  scale_fill_brewer(palette="Set1") +
  coord_cartesian(ylim=c(.3,.8)) +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
    axis.title=element_text(size=12,face = "bold"),
    axis.text=element_text(size=14),
    legend.position="right",
    strip.text.x = element_text(size = 12))
p.RR.3
```



## Analysis: RR by liquid, block, and age group

```
# Monetary Reward
m.RR.motblock <-glmer(formula = subRewarded ~ liqCode * blockCode * groupCode +
                      moneyCode + (1+moneyCode+liqCode|subID),
                      data = incentive, family="binomial")
summary(m.RR.motblock)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode * blockCode * groupCode + moneyCode + (1 +
## moneyCode + liqCode | subID)
## Data: incentive
##
##      AIC      BIC   logLik deviance df.resid
## 35654.5 35778.2 -17812.3 35624.5    28209
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.9435 -1.0556  0.5710  0.7645  5.0085
##
## Random effects:
## Groups Name          Variance Std.Dev. Corr
## subID (Intercept) 0.24656  0.4965
## moneyCode 0.02148  0.1466  -0.20
## liqCode 0.09213  0.3035  -0.30  0.47
## Number of obs: 28224, groups: subID, 98
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.66229    0.07278   9.100 < 2e-16 ***
## liqCode         0.25012    0.04784   5.229 1.71e-07 ***
## blockCode       0.06688    0.02725   2.454 0.014114 *
## groupCode      -0.36059    0.11395  -3.165 0.001553 **
## moneyCode       0.17849    0.02165   8.245 < 2e-16 ***
## liqCode:blockCode 0.16939    0.04520   3.748 0.000178 ***
## liqCode:groupCode -0.19524    0.07321  -2.667 0.007655 **
## blockCode:groupCode 0.19275    0.03944   4.887 1.02e-06 ***
## liqCode:blockCode:groupCode -0.25673    0.06752  -3.803 0.000143 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod blkCd  grpCd monyCd lqCd:bc lqCd:gC blcC:C
## liqCode      -0.238
## blockCode     0.004  0.023
## groupCode    -0.702  0.156 -0.002
## moneyCode    -0.097  0.202 -0.001  0.012
## lqCd:blkCd   0.010 -0.001 -0.046 -0.016  0.002
## liqCd:grpCd  0.160 -0.691 -0.030 -0.232  0.011  0.015
## blkCd:grpCd -0.005  0.001 -0.688  0.009  0.003  0.027 -0.024
## lqCd:blcC:C  0.006  0.008  0.032 -0.021 -0.005 -0.671 -0.005 -0.070
```

```
# OA
m.RR.motblock.OA <-glmer(formula = subRewarded ~ liqCode * blockCode + moneyCode +
                          (1+moneyCode+liqCode|subID),
                          data = subset(incentive,group=="OA"), family="binomial")
summary(m.RR.motblock.OA)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode * blockCode + moneyCode + (1 + moneyCode +
## liqCode | subID)
## Data: subset(incentive, group == "OA")
##
##      AIC      BIC    logLik deviance df.resid
## 16589.7 16671.6 -8283.8 16567.7    12661
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.138 -1.048  0.619  0.822  2.131
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.222844 0.47206
## moneyCode 0.006331 0.07957 -0.34
## liqCode 0.030644 0.17505 0.12 -0.14
## Number of obs: 12672, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.32959    0.07361  4.478 7.55e-06 ***
## liqCode         0.01075    0.03500  0.307 0.75867
## blockCode       0.27235    0.02712 10.042 < 2e-16 ***
## moneyCode       0.08321    0.02579  3.226 0.00126 **
## liqCode:blockCode -0.05635    0.04704 -1.198 0.23100
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod blkCd monyCd
## liqCode      0.088
## blockCode    0.008 -0.052
## moneyCode   -0.152 -0.050  0.004
## lqCd:blkCd -0.028  0.042 -0.025 -0.005
```

```
# Money & Liquid Effects in Block 1
m.RR.motblock.OA.b1 <-glmer(formula = subRewarded ~ liqCode + moneyCode +
                          (1+moneyCode+liqCode|subID),
                          data = subset(incentive,group=="OA" & block==1), family="binomial")
summary(m.RR.motblock.OA.b1)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode + moneyCode + (1 + moneyCode + liqCode |
## subID)
```

```

## Data: subset(incentive, group == "OA" & block == 1)
##
##      AIC      BIC   logLik deviance df.resid
##  5722.5   5779.6  -2852.2   5704.5     4215
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -1.6734 -0.9748  0.5976  0.9310  1.8984
##
## Random effects:
##  Groups Name      Variance Std.Dev. Corr
##  subID  (Intercept) 0.164805 0.40596
##          moneyCode  0.003403 0.05833  0.18
##          liqCode    0.077317 0.27806  0.10 0.10
## Number of obs: 4224, groups:  subID, 44
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.01995    0.08068   0.247   0.8047
## liqCode      0.18550    0.10823   1.714   0.0865 .
## moneyCode    0.03127    0.03975   0.787   0.4315
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod
## liqCode      0.119
## moneyCode    0.041  0.005
##
# Money & Liquid Effects in Block 2
m.RR.motblock.OA.b2 <-glmer(formula = subRewarded ~ liqCode + moneyCode +
                           (1+moneyCode+liqCode|subID),
                           data = subset(incentive,group=="OA" & block==2), family="binomial")
summary(m.RR.motblock.OA.b2)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode + moneyCode + (1 + moneyCode + liqCode |
##          subID)
## Data: subset(incentive, group == "OA" & block == 2)
##
##      AIC      BIC   logLik deviance df.resid
##  5542.0   5599.2  -2762.0   5524.0     4215
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.2945 -1.0912  0.6454  0.7949  1.8075
##
## Random effects:
##  Groups Name      Variance Std.Dev. Corr
##  subID  (Intercept) 0.27706  0.52636
##          moneyCode  0.00557  0.07463  -1.00
##          liqCode    0.03000  0.17322   1.00 -1.00
## Number of obs: 4224, groups:  subID, 44

```

```
##
## Fixed effects:
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.38457    0.09071   4.239 2.24e-05 ***
## liqCode      -0.16117    0.10069  -1.601   0.109
## moneyCode     0.03905    0.04127   0.946   0.344
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr) liqCod
## liqCode     0.529
## moneyCode  -0.243 -0.076

# Money & Effects in Block 3
m.RR.motblock.OA.b3 <-glmer(formula = subRewarded ~ liqCode + moneyCode +
                           (1+moneyCode+liqCode|subID),
                           data = subset(incentive,group=="OA" & block==3), family="binomial")
summary(m.RR.motblock.OA.b3)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode + moneyCode + (1 + moneyCode + liqCode |
## subID)
## Data: subset(incentive, group == "OA" & block == 3)
##
##      AIC      BIC   logLik deviance df.resid
##  5371.0   5428.1  -2676.5   5353.0     4215
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.3554 -1.1029  0.5802  0.7455  1.5151
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.263441 0.51326
## moneyCode 0.004076 0.06385 -0.26
## liqCode 0.057171 0.23910 -0.06 -0.95
## Number of obs: 4224, groups: subID, 44
##
## Fixed effects:
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.58050    0.08988   6.459 1.05e-10 ***
## liqCode       0.03633    0.11869   0.306   0.76
## moneyCode     0.18430    0.04200   4.388 1.14e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr) liqCod
## liqCode    -0.169
## moneyCode  -0.035 -0.076
```

```
# YA
m.RR.motblock.YA <-glmer(formula = subRewarded ~ liqCode * blockCode + moneyCode +
                          (1+moneyCode+liqCode|subID),
                          data = subset(incentive,group=="YA"), family="binomial")
summary(m.RR.motblock.YA)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode * blockCode + moneyCode + (1 + moneyCode +
## liqCode | subID)
## Data: subset(incentive, group == "YA")
##
##      AIC      BIC   logLik deviance df.resid
## 19038.8 19122.9 -9508.4 19016.8    15541
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -4.1708 -1.0655  0.5390  0.7158  6.1846
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.27079  0.5204
## moneyCode 0.02136  0.1462  -0.15
## liqCode 0.14841  0.3852  -0.48  0.62
## Number of obs: 15552, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.63788    0.07307  8.729 < 2e-16 ***
## liqCode         0.28885    0.05694  5.073 3.93e-07 ***
## blockCode       0.06885    0.02783  2.474 0.013347 *
## moneyCode       0.26292    0.02954  8.901 < 2e-16 ***
## liqCode:blockCode 0.17881    0.04614  3.876 0.000106 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod blkCd monyCd
## liqCode      -0.425
## blockCode     0.003  0.004
## moneyCode    -0.085  0.396 -0.002
## lqCd:blkCd   -0.004  0.013 -0.069  0.004
```

```
# block 1
m.RR.motblock.YA.b1 <-glmer(formula = subRewarded ~ liqCode + moneyCode +
                             (1+moneyCode+liqCode|subID),
                             data = subset(incentive,group=="YA" & block==1), family="binomial")
summary(m.RR.motblock.YA.b1)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode + moneyCode + (1 + moneyCode + liqCode |
## subID)
```

```

## Data: subset(incentive, group == "YA" & block == 1)
##
##      AIC      BIC   logLik deviance df.resid
##  6698.8   6757.7 -3340.4   6680.8     5175
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.0039 -1.1322  0.6226  0.7447  1.4254
##
## Random effects:
##   Groups Name      Variance Std.Dev. Corr
##   subID  (Intercept) 0.11870  0.3445
##           moneyCode  0.04040  0.2010  -0.10
##           liqCode    0.09093  0.3016   0.15  0.97
## Number of obs: 5184, groups:  subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   0.59193    0.06510   9.093 < 2e-16 ***
## liqCode       -0.09227    0.08689  -1.062   0.288
## moneyCode     0.20272    0.04558   4.447  8.7e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod
## liqCode      0.142
## moneyCode   -0.008  0.293
# block 2
m.RR.motblock.YA.b2 <-glmer(formula = subRewarded ~ liqCode + moneyCode +
                           (1+moneyCode+liqCode|subID),
                           data = subset(incentive,group=="YA" & block==2), family="binomial")

## Warning in checkConv(attr(opt, "derivs"), opt$par, ctrl = control
## $checkConv, : Model failed to converge with max|grad| = 0.00225551 (tol =
## 0.001, component 1)

summary(m.RR.motblock.YA.b2)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode + moneyCode + (1 + moneyCode + liqCode |
##          subID)
## Data: subset(incentive, group == "YA" & block == 2)
##
##      AIC      BIC   logLik deviance df.resid
##  6205.0   6264.0 -3093.5   6187.0     5175
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.5748 -1.0683  0.5376  0.6636  5.6509
##
## Random effects:

```

```

## Groups Name      Variance Std.Dev. Corr
## subID (Intercept) 0.15184  0.3897
##      moneyCode    0.03717  0.1928  -0.56
##      liqCode      0.61801  0.7861  -0.99  0.42
## Number of obs: 5184, groups: subID, 54
##
## Fixed effects:
##      Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.76017    0.08707   8.730 < 2e-16 ***
## liqCode      0.30629    0.12591   2.433  0.015 *
## moneyCode    0.19350    0.04709   4.109 3.97e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr) liqCod
## liqCode  -0.455
## moneyCode -0.198  0.176
## convergence code: 0
## Model failed to converge with max|grad| = 0.00225551 (tol = 0.001, component 1)

# block 3
m.RR.motblock.YA.b3 <-glmer(formula = subRewarded ~ liqCode + moneyCode +
                           (1+moneyCode+liqCode|subID),
                           data = subset(incentive,group=="YA" & block==3), family="binomial")
summary(m.RR.motblock.YA.b3)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula: subRewarded ~ liqCode + moneyCode + (1 + moneyCode + liqCode |
##      subID)
## Data: subset(incentive, group == "YA" & block == 3)
##
##      AIC      BIC    logLik deviance df.resid
##  6100.7   6159.7  -3041.4   6082.7     5175
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4693 -0.9137  0.4960  0.6602  3.5257
##
## Random effects:
## Groups Name      Variance Std.Dev. Corr
## subID (Intercept) 0.46513  0.6820
##      moneyCode    0.09152  0.3025  -0.51
##      liqCode      0.01677  0.1295  -0.97  0.69
## Number of obs: 5184, groups: subID, 54
##
## Fixed effects:
##      Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.70623    0.10245   6.893 5.45e-12 ***
## liqCode      0.42913    0.12582   3.411 0.000648 ***
## moneyCode    0.38989    0.05752   6.779 1.21e-11 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```



```
##
## Correlation of Fixed Effects:
##      (Intr) liqCod
## liqCode  -0.189
## moneyCode -0.302  0.136
```

## Analysis comparing OA performance in block 3 vs YA performance in block 1

```
# OA block 3 vs YA block 1
```

```
t.test(x = subset(liq.means, group=="OA" & block==3)$meanRR,
       y = subset(liq.means, group=="YA" & block==1)$meanRR,
       paired=FALSE)
```

```
##
## Welch Two Sample t-test
##
## data:  subset(liq.means, group == "OA" & block == 3)$meanRR and subset(liq.means, group == "YA" & bl
## t = 0.20309, df = 83.311, p-value = 0.8396
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  -0.04417837  0.05422678
## sample estimates:
## mean of x mean of y
## 0.6323390 0.6273148
```

## Reinforcement Effects Model: Trialwise Estimated Values

```
# Define constant parameters
L = 0.5
val_init = 0.5

# Loop over subjects and add estimated value based on previously rewarded trial
SUBJECTS = unique(incentive$subID)
#SUBJECTS = 1
incentive$value_est=999
for (subj.id in SUBJECTS) {
  # filter data for subject
  tmp.data<-filter(incentive,subID==subj.id)

  # remove values_subj vector
  if (exists("values_subj")) {rm(values_subj)}
  if (exists("values_subj_param")) {rm(values_subj_param)}

  # loop over blocks (1,2,3)
  BLOCKS=as.numeric(unique(tmp.data$block))
  for (b in BLOCKS) {
    # filter data for block
    tmp.data.block<-filter(tmp.data,block==b)
    # loop over trials within each block
    values = vector(mode = "numeric", length=96)
    values_param = vector(mode = "numeric", length=96)
    for (t in tmp.data.block$Trial) {
      # calculate estimated value per trial based on reinforcement/previous reward
      # non-parametric version: based only previous reward only
      # parametric version: based on previous reward & monetary value
      if (t==1) {
        values[t]=val_init
        values_param[t]=val_init
      } else {
        # nonparametric
        values[t]=((1-L)*values[t-1]) + L*as.numeric(tmp.data.block$subRewarded[t-1])
        # parametric
        w = tmp.data.block$moneyweight[t-1]
        values_param[t]=((1-L)*values_param[t-1]) +
          L*as.numeric(w*tmp.data.block$subRewarded[t-1])
      }
    }

    # end for loop over trials

    # concatenate the values from blocks for each subject
    if (exists("values_subj")) {values_subj<-c(values_subj,values)}
    } else {values_subj<-values}

    # concatenate the values from blocks for each subject: parametric
    if (exists("values_subj_param")) {values_subj_param<-c(values_subj_param,values_param)}
    } else {values_subj_param<-values_param}

  } # end for loop over blocks
```

```

# identify indices for where in data frame to enter value estimates
ix<-which(incentive$subID==subj.id)
# add value estimates to the data frame
incentive[ix,"value_est"]=round(values_subj,4)
incentive[ix,"value_est_param"]=round(values_subj_param,4)

} # end for loop over subjects

```

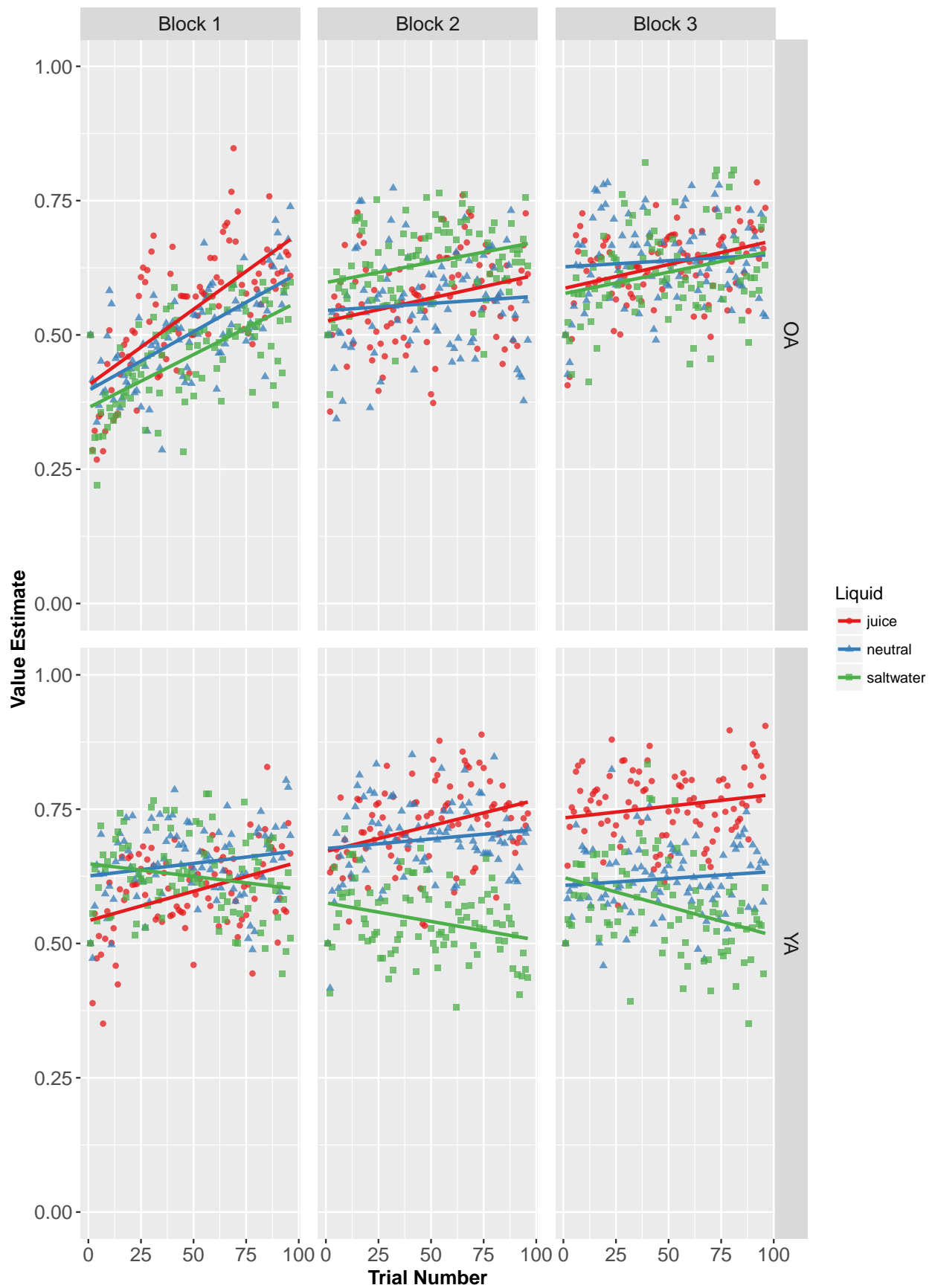
## Plotting Group Averaged Value Estimates

```

data.group.value.estimates <- incentive %>% group_by(Trial,group,liquid,block) %>%
  summarise(n=n(),value_est_mean = mean(value_est),
            value_est_param_mean = mean(value_est_param)) %>%
  arrange(block,liquid,group,Trial)
levels(data.group.value.estimates$block)[levels(data.group.value.estimates$block)=="1"] <- "Block 1"
levels(data.group.value.estimates$block)[levels(data.group.value.estimates$block)=="2"] <- "Block 2"
levels(data.group.value.estimates$block)[levels(data.group.value.estimates$block)=="3"] <- "Block 3"
names(data.group.value.estimates)[3]<-"Liquid"

# Plotting group value estimates
p.val.1<-ggplot(data = data.group.value.estimates,
               mapping=aes(x=Trial,y=value_est_mean,
                           color=Liquid, shape=Liquid)) +
  geom_point(alpha=0.75) +
  geom_smooth(method="lm", se=FALSE) +
  coord_cartesian(ylim=c(0,1), xlim=c(1,96)) +
  facet_grid(group~block) +
  xlab("Trial Number") + ylab("Value Estimate") +
  #labs(shape = "Liquid") +
  scale_color_brewer(palette="Set1") +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=12),
        #legend.position="none",
        strip.text.x = element_text(size = 12),
        strip.text.y = element_text(size = 12),
        panel.spacing = unit(.75, "lines"))
p.val.1

```



## Generalized Linear Mixed Model

Logistic regression of Estimated Reward Value and Experimental Conditions (money, liquid, blocks), as well as group (older adults vs younger adults).

```
# Reward Rate predicted by estimated value (reinforcement)
m.val.1 <-glmer(formula = subRewarded ~ value_est + (1|subID),
                data = incentive, family=binomial)
summary(m.val.1)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial (logit)
## Formula: subRewarded ~ value_est + (1 | subID)
## Data: incentive
##
##      AIC      BIC   logLik deviance df.resid
## 36233.2 36258.0 -18113.6 36227.2    28221
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.2937 -1.1009  0.6028  0.7804  1.6912
##
## Random effects:
## Groups Name      Variance Std.Dev.
## subID (Intercept) 0.2101   0.4583
## Number of obs: 28224, groups: subID, 98
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.23437    0.05515   4.250 2.14e-05 ***
## value_est    0.41763    0.04484   9.314 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr)
## value_est -0.492
```

```
# money, liquid, age group effects, including estimated value
m.val.2 <-glmer(formula = subRewarded ~ moneyCode*liqCode + value_est*blockCode*groupCode +
                value_est:liqCode +
                (1+moneyCode+liqCode|subID),
                data = incentive, family=binomial)
summary(m.val.2)
```

```
## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial (logit)
## Formula:
## subRewarded ~ moneyCode * liqCode + value_est * blockCode * groupCode +
## value_est:liqCode + (1 + moneyCode + liqCode | subID)
## Data: incentive
##
##      AIC      BIC   logLik deviance df.resid
```

```

## 35667.8 35816.2 -17815.9 35631.8 28206
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.6985 -1.0564  0.5669  0.7640  5.1816
##
## Random effects:
##      Groups Name      Variance Std.Dev. Corr
## subID (Intercept) 0.24843  0.4984
##      moneyCode   0.02138  0.1462  -0.23
##      liqCode     0.10239  0.3200  -0.29  0.57
## Number of obs: 28224, groups: subID, 98
##
## Fixed effects:
##
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.63727    0.08530   7.471 7.95e-14 ***
## moneyCode         0.17872    0.02163   8.264 < 2e-16 ***
## liqCode           0.16828    0.05052   3.331 0.000866 ***
## value_est         0.08312    0.06643   1.251 0.210893
## blockCode         0.05506    0.05583   0.986 0.324012
## groupCode        -0.49130    0.12947  -3.795 0.000148 ***
## moneyCode:liqCode -0.03807    0.01934  -1.969 0.048976 *
## value_est:blockCode 0.02267    0.07711   0.294 0.768734
## value_est:groupCode 0.12709    0.09364   1.357 0.174691
## blockCode:groupCode 0.22545    0.07758   2.906 0.003661 **
## liqCode:value_est -0.01797    0.05790  -0.310 0.756239
## value_est:blockCode:groupCode -0.09617    0.11077  -0.868 0.385283
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##      (Intr) monyCd liqCod val_st blkCd gropCd mnyC:C vl_st:bC
## moneyCode  -0.097
## liqCode     -0.070  0.251
## value_est   -0.511  0.003 -0.086
## blockCode   0.074 -0.003  0.053 -0.084
## groupCode  -0.714  0.010 -0.005  0.341 -0.068
## monyCd:liqCd -0.009  0.017  0.002  0.002  0.000  0.012
## vl_st:blkC  -0.058  0.003 -0.053  0.075 -0.870  0.049 -0.003
## vl_st:grpCd  0.356 -0.002  0.059 -0.698  0.058 -0.451  0.005 -0.050
## blkCd:grpC  -0.053  0.011 -0.059  0.062 -0.721  0.090 -0.002  0.627
## liqCd:vl_st -0.021  0.003 -0.699  0.059 -0.032 -0.015  0.010  0.028
## vl_st:blC:C  0.041 -0.012  0.058 -0.053  0.607 -0.065  0.002 -0.697
##      vl_st:gC blcC:C lqCd:_
## moneyCode
## liqCode
## value_est
## blockCode
## groupCode
## monyCd:liqCd
## vl_st:blkC
## vl_st:grpCd
## blkCd:grpC -0.107
## liqCd:vl_st -0.043  0.040

```

```

## vl_st:blc:C 0.069 -0.857 -0.043
# younger adults value estimate
m.val.2a <-glmer(formula = subRewarded ~ liqCode*moneyCode + blockCode*value_est +
                  value_est:liqCode + (1+moneyCode+liqCode|subID),
                  data = filter(incentive,group=="YA" & liquid!="neutral"), family=binomial)
summary(m.val.2a)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula:
## subRewarded ~ liqCode * moneyCode + blockCode * value_est + value_est:liqCode +
## (1 + moneyCode + liqCode | subID)
## Data: filter(incentive, group == "YA" & liquid != "neutral")
##
##      AIC      BIC   logLik deviance df.resid
## 12667.9 12769.3 -6319.9 12639.9    10354
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.6251 -1.0473  0.5331  0.7220  4.6647
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.29392  0.5421
## moneyCode 0.02011  0.1418 -0.02
## liqCode 0.14271  0.3778 -0.58 0.67
## Number of obs: 10368, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.621990   0.094514   6.581 4.67e-11 ***
## liqCode         0.327795   0.078316   4.186 2.85e-05 ***
## moneyCode       0.280133   0.033133   8.455 < 2e-16 ***
## blockCode       0.242138   0.092526   2.617 0.00887 **
## value_est      -0.004148   0.084352  -0.049 0.96078
## liqCode:moneyCode -0.059937   0.026798  -2.237 0.02531 *
## blockCode:value_est -0.036371   0.099485  -0.366 0.71467
## liqCode:value_est -0.085001   0.084046  -1.011 0.31185
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod monyCd blkCd val_st lqCd:C blcC:_
## liqCode      -0.174
## moneyCode    -0.003  0.261
## blockCode     0.089  0.074 -0.004
## value_est    -0.576 -0.139  0.006 -0.126
## liqCd:monyCd  0.007  0.014  0.031 -0.005  0.001
## blkCd:vl_s   -0.093 -0.118  0.002 -0.719  0.120 -0.011
## liqCd:vl_st -0.114 -0.694  0.001 -0.062  0.103  0.003  0.095
# older adults value estimate
m.val.2b <-glmer(formula = subRewarded ~ liqCode*moneyCode + blockCode*value_est +

```

```

value_est:liqCode + (1+moneyCode+liqCode|subID),
data = filter(incentive,group=="0A"), family=binomial)
summary(m.val.2b)

## Generalized linear mixed model fit by maximum likelihood (Laplace
## Approximation) [glmerMod]
## Family: binomial ( logit )
## Formula:
## subRewarded ~ liqCode * moneyCode + blockCode * value_est + value_est:liqCode +
## (1 + moneyCode + liqCode | subID)
## Data: filter(incentive, group == "0A")
##
##      AIC      BIC    logLik deviance df.resid
## 16581.4 16685.6 -8276.7 16553.4    12658
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.0904 -1.0467  0.6185  0.8178  2.1509
##
## Random effects:
## Groups Name             Variance Std.Dev. Corr
## subID (Intercept) 0.200844 0.44816
## moneyCode 0.006291 0.07931 -0.37
## liqCode 0.022389 0.14963 0.07 -0.24
## Number of obs: 12672, groups: subID, 44
##
## Fixed effects:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)    0.19314    0.08026   2.407 0.016104 *
## liqCode        -0.04322    0.05741  -0.753 0.451605
## moneyCode       0.08424    0.02579   3.267 0.001088 **
## blockCode       0.30862    0.05214   5.919 3.23e-09 ***
## value_est       0.23903    0.06770   3.531 0.000414 ***
## liqCode:moneyCode -0.01155    0.02794  -0.413 0.679248
## blockCode:value_est -0.08901    0.07844  -1.135 0.256504
## liqCode:value_est  0.09454    0.08276   1.142 0.253309
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod monyCd blkCd val_st lqCd:C blcC:_
## liqCode      0.022
## moneyCode    -0.141 -0.056
## blockCode     0.096 -0.059  0.024
## value_est    -0.485  0.005  0.001 -0.126
## liqCd:monyCd -0.007 -0.005  0.004 -0.004  0.014
## blkCd:vl_s   -0.068  0.054 -0.026 -0.857  0.063  0.001
## liqCd:vl_st  0.002 -0.828  0.013  0.041 -0.007  0.009 -0.050

```



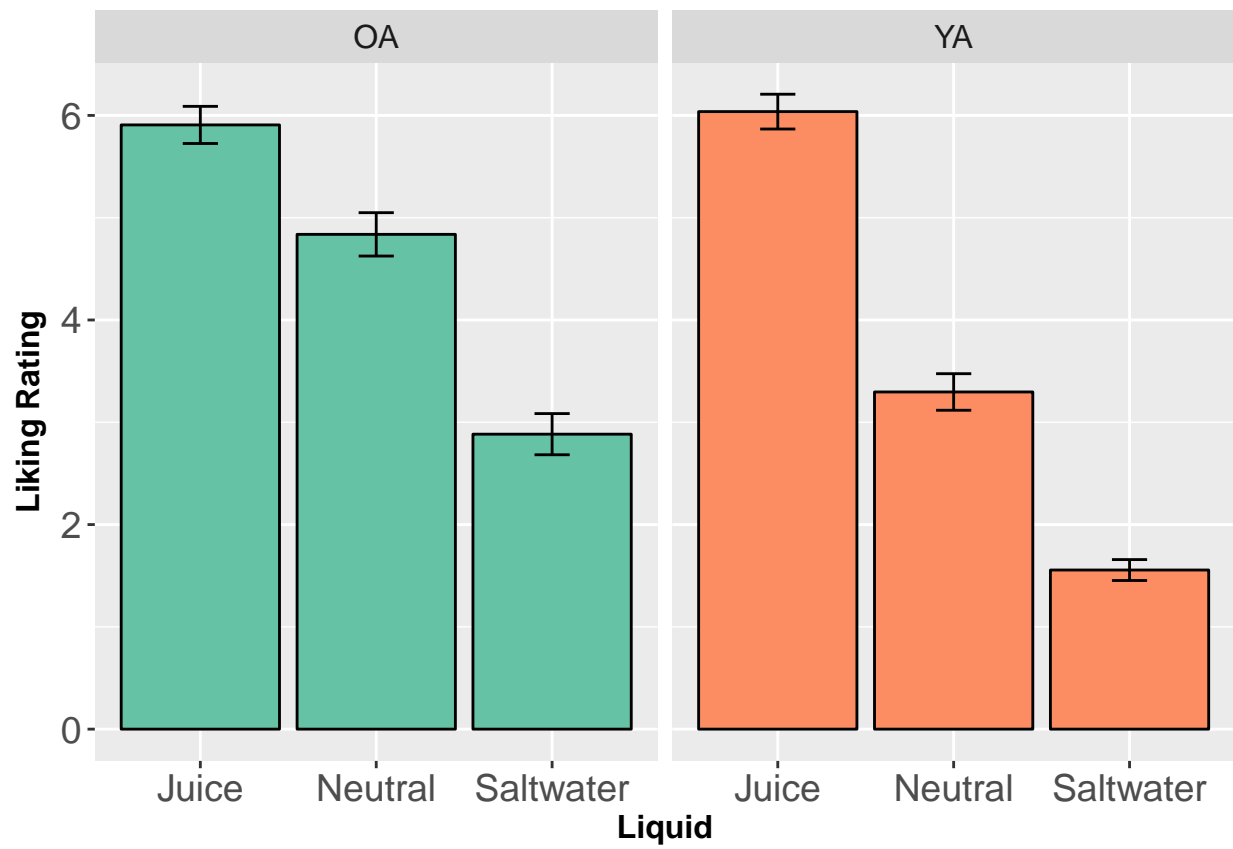
# Self-Report Ratings

## Formatting Data

```
data.selfreport<-read.csv(data.path.selfreport) %>%  
  filter(include=="yes")
```

## Plot of Self-Report Liking Ratings per Liquid

```
# formatting the liking data from wide to long form  
data.liking<-data.selfreport %>%  
  select(subID, group, Juice=juice_like, Neutral=neut_like, Saltwater=salt_like) %>%  
  gather(key = liquid, value = rating, Juice:Saltwater, factor_key=TRUE) %>%  
  mutate(liqCode = factor(liquid, levels = c("Saltwater","Neutral","Juice"), labels=c(-1,0,1)),  
         groupCode = factor(group, levels=c("YA","OA"), labels=c(0,1)))  
data.liking$liqCode<-as.numeric(levels(data.liking$liqCode)[data.liking$liqCode])  
data.liking$groupCode<-as.numeric(levels(data.liking$groupCode)[data.liking$groupCode])  
  
selfreport.sum<-summarySEwithin2(data = data.liking, measurevar = "rating",  
                                withinvars = c("liquid"), betweenvars = "group",  
                                idvar = "subID")  
p.rating.1<-ggplot(data = selfreport.sum, aes(x = liquid, y = rating, fill=group)) +  
  geom_bar(stat = "identity", color="black") +  
  geom_errorbar(mapping = aes(ymin=rating-se, ymax=rating+se), width=.2) +  
  xlab("Liquid") + ylab("Liking Rating") +  
  scale_x_discrete(labels=c("juice_like" = "Juice",  
                           "neut_like" = "Neutral",  
                           "salt_like" = "Saltwater")) +  
  
  facet_grid(.~group) +  
  scale_fill_brewer(palette = "Set2") +  
  #coord_cartesian(ylim=c(1,7)) +  
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),  
        axis.title=element_text(size=12,face = "bold"),  
        axis.text=element_text(size=14),  
        legend.position="none",  
        strip.text.x = element_text(size = 12))  
p.rating.1
```



## Stats on Liking Ratings

```
m.selfreport.1<-lmer(formula = rating ~ liqCode*groupCode + (1|subID), data = data.liking)
summary(m.selfreport.1)
```

```
## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * groupCode + (1 | subID)
## Data: data.liking
##
## REML criterion at convergence: 972.2
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.2081 -0.5144 -0.0159  0.6245  2.6332
##
## Random effects:
##  Groups   Name                Variance Std.Dev.
## subID    (Intercept)  0.1007   0.3174
## Residual                    1.5272   1.2358
## Number of obs: 291, groups: subID, 97
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    3.6296    0.1063  95.0000   34.156 < 2e-16 ***
## liqCode         2.2407    0.1189 192.0000   18.843 < 2e-16 ***
## groupCode       0.9130    0.1596  95.0000    5.720 1.23e-07 ***
## liqCode:groupCode -0.7291    0.1786 192.0000   -4.082 6.54e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod gropCd
## liqCode      0.000
## groupCode   -0.666  0.000
## liqCd:grpCd  0.000 -0.666  0.000
```

# OA

```
m.selfreport.1.OA<-lmer(formula = rating ~ liqCode + (1|subID),
                        data = subset(data.liking, group=="OA"))
summary(m.selfreport.1.OA)
```

```
## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode + (1 | subID)
## Data: subset(data.liking, group == "OA")
##
## REML criterion at convergence: 458.6
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.92532 -0.71254  0.01504  0.68307  1.94733
##
## Random effects:
```

```

## Groups Name Variance Std.Dev.
## subID (Intercept) 0.3186 0.5644
## Residual 1.7430 1.3202
## Number of obs: 129, groups: subID, 43
##
## Fixed effects:
## Estimate Std. Error df t value Pr(>|t|)
## (Intercept) 4.5426 0.1446 42.0000 31.41 <2e-16 ***
## liqCode 1.5116 0.1424 85.0000 10.62 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr)
## liqCode 0.000

# OA JuN
t.test(x = subset(data.liking, liquid=="Juice" & group=="OA")$rating,
       y = subset(data.liking, liquid=="Neutral" & group=="OA")$rating,
       paired = TRUE)

##
## Paired t-test
##
## data: subset(data.liking, liquid == "Juice" & group == "OA")$rating and subset(data.liking, liquid == "Neutral" & group == "OA")$rating
## t = 3.8589, df = 42, p-value = 0.000386
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.510319 1.629216
## sample estimates:
## mean of the differences
## 1.069767

# OA NuS
t.test(x = subset(data.liking, liquid=="Neutral" & group=="OA")$rating,
       y = subset(data.liking, liquid=="Saltwater" & group=="OA")$rating,
       paired = TRUE)

##
## Paired t-test
##
## data: subset(data.liking, liquid == "Neutral" & group == "OA")$rating and subset(data.liking, liquid == "Saltwater" & group == "OA")$rating
## t = 6.4452, df = 42, p-value = 9.085e-08
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 1.341825 2.565151
## sample estimates:
## mean of the differences
## 1.953488

# YA
m.selfreport.1.YA<-lmer(formula = rating ~ liqCode + (1|subID),
                       data = subset(data.liking, group=="YA"))
summary(m.selfreport.1.YA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations

```

```
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode + (1 | subID)
## Data: subset(data.liking, group == "YA")
##
## REML criterion at convergence: 503.9
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4150 -0.5555  0.1144  0.5392  2.9738
##
## Random effects:
##   Groups   Name      Variance Std.Dev.
##   subID    (Intercept) 1.599e-19 3.998e-10
##   Residual                1.284e+00 1.133e+00
## Number of obs: 162, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   3.62963    0.08904 160.00000   40.76 <2e-16 ***
## liqCode        2.24074    0.10906 160.00000   20.55 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr)
## liqCode 0.000
```

```
# YA JuN
t.test(x = subset(data.liking, liquid=="Juice" & group=="YA")$rating,
       y = subset(data.liking, liquid=="Neutral" & group=="YA")$rating,
       paired = TRUE)
```

```
##
## Paired t-test
##
## data: subset(data.liking, liquid == "Juice" & group == "YA")$rating and subset(data.liking, liquid == "Neutral" & group == "YA")$rating
## t = 10.061, df = 53, p-value = 6.791e-14
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  2.194369 3.287112
## sample estimates:
## mean of the differences
##                2.740741
```

```
# YA NuS
t.test(x = subset(data.liking, liquid=="Neutral" & group=="YA")$rating,
       y = subset(data.liking, liquid=="Saltwater" & group=="YA")$rating,
       paired = TRUE)
```

```
##
## Paired t-test
##
## data: subset(data.liking, liquid == "Neutral" & group == "YA")$rating and subset(data.liking, liquid == "Saltwater" & group == "YA")$rating
## t = 9.0294, df = 53, p-value = 2.637e-12
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
```

```

## 1.354061 2.127421
## sample estimates:
## mean of the differences
## 1.740741

# OA vs YA
# Juice
t.test(x = subset(data.liking, liquid=="Juice" & group=="OA")$rating,
       y = subset(data.liking, liquid=="Juice" & group=="YA")$rating)

##
## Welch Two Sample t-test
##
## data: subset(data.liking, liquid == "Juice" & group == "OA")$rating and subset(data.liking, liquid == "Juice" & group == "YA")$rating
## t = -0.53693, df = 84.27, p-value = 0.5927
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.6117368 0.3516163
## sample estimates:
## mean of x mean of y
## 5.906977 6.037037

# Neutral
t.test(x = subset(data.liking, liquid=="Neutral" & group=="OA")$rating,
       y = subset(data.liking, liquid=="Neutral" & group=="YA")$rating)

##
## Welch Two Sample t-test
##
## data: subset(data.liking, liquid == "Neutral" & group == "OA")$rating and subset(data.liking, liquid == "Neutral" & group == "YA")$rating
## t = 5.1571, df = 84.267, p-value = 1.636e-06
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.9467597 2.1350663
## sample estimates:
## mean of x mean of y
## 4.837209 3.296296

# Saltwater
t.test(x = subset(data.liking, liquid=="Saltwater" & group=="OA")$rating,
       y = subset(data.liking, liquid=="Saltwater" & group=="YA")$rating)

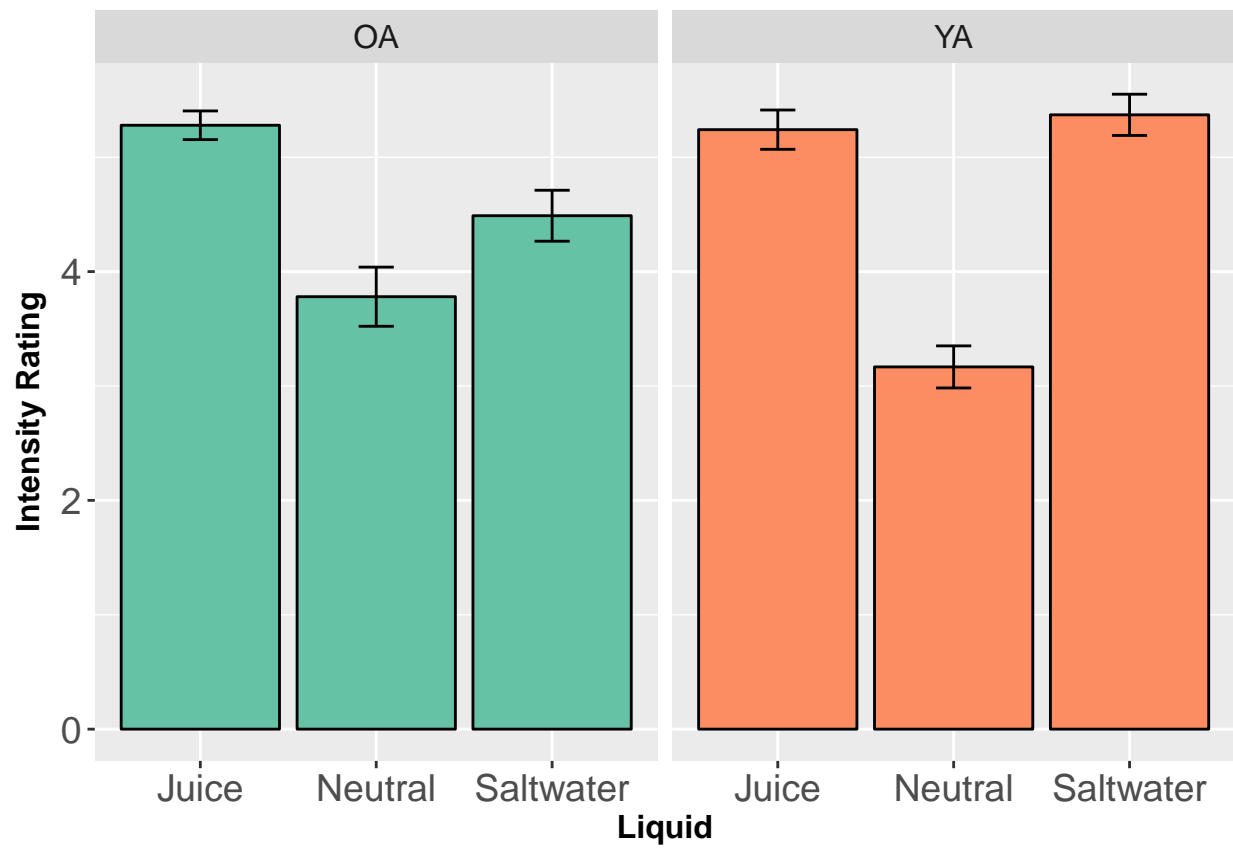
##
## Welch Two Sample t-test
##
## data: subset(data.liking, liquid == "Saltwater" & group == "OA")$rating and subset(data.liking, liquid == "Saltwater" & group == "YA")$rating
## t = 5.3164, df = 62.267, p-value = 1.519e-06
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## 0.8288196 1.8275111
## sample estimates:
## mean of x mean of y
## 2.883721 1.555556

```

## Plot of Self-Report Intensity Ratings per Liquid

```
# formatting the intensity data from wide to long form
data.intensity<-data.selfreport %>%
  select(subID, group, Juice=juice_intense, Neutral=neut_intense, Saltwater=salt_intense) %>%
  gather(key = liquid, value = rating, Juice:Saltwater, factor_key=TRUE) %>%
  mutate(liqCode = factor(liquid, levels = c("Saltwater","Neutral","Juice"), labels=c(-1,0,1)),
         groupCode = factor(group, levels=c("YA","OA"), labels=c(0,1)))
data.intensity$liqCode<-as.numeric(levels(data.intensity$liqCode)[data.intensity$liqCode])
data.intensity$groupCode<-as.numeric(levels(data.intensity$groupCode)[data.intensity$groupCode])

selfreport.sum<-summarySEwithin2(data = data.intensity, measurevar = "rating",
                                withinvars = c("liquid"), betweenvars = "group",
                                idvar = "subID")
p.rating.2<-ggplot(data = selfreport.sum, aes(x = liquid, y = rating, fill=group)) +
  geom_bar(stat = "identity", color="black") +
  geom_errorbar(mapping = aes(ymin=rating-se, ymax=rating+se), width=.2) +
  xlab("Liquid") + ylab("Intensity Rating") +
  scale_x_discrete(labels=c("juice_intense" = "Juice",
                           "neut_intense" = "Neutral",
                           "salt_intense" = "Saltwater")) +
  facet_grid(.~group) +
  scale_fill_brewer(palette = "Set2") +
  #coord_cartesian(ylim=c(1,7)) +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position="none",
        strip.text.x = element_text(size = 12))
p.rating.2
```





## Stats on Intensity Ratings

```
m.selfreport.2<-lmer(formula = rating ~ liqCode*groupCode + (1|subID),
                     data = data.intensity)
summary(m.selfreport.2)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * groupCode + (1 | subID)
## Data: data.intensity
##
## REML criterion at convergence: 1112.9
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.2181 -0.5597  0.2078  0.8142  1.7392
##
## Random effects:
## Groups Name Variance Std.Dev.
## subID (Intercept) 9.798e-16 3.130e-08
## Residual 2.719e+00 1.649e+00
## Number of obs: 289, groups: subID, 97
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    4.59259    0.12955 285.00000   35.451  <2e-16 ***
## liqCode        -0.06481    0.15866 285.00000   -0.409   0.6832
## groupCode      -0.06503    0.19543 285.00000   -0.333   0.7395
## liqCode:groupCode 0.46016    0.23830 285.00000    1.931   0.0545 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod gropCd
## liqCode      0.000
## groupCode   -0.663  0.000
## liqCd:grpCd  0.000 -0.666  0.000

# OA
m.selfreport.2.OA<-lmer(formula = rating ~ liqCode + (1|subID),
                       data = subset(data.intensity, group=="OA"))
summary(m.selfreport.2.OA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode + (1 | subID)
## Data: subset(data.intensity, group == "OA")
##
## REML criterion at convergence: 458.5
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.41722 -0.36150  0.05283  0.73806  1.96512
##
```

```

## Random effects:
##   Groups   Name      Variance Std.Dev.
##   subID    (Intercept) 0.00      0.000
##   Residual                2.13      1.459
## Number of obs: 127, groups:  subID, 43
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   4.5276     0.1295 125.0000  34.963   <2e-16 ***
## liqCode        0.3953     0.1574 125.0000   2.512   0.0133 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr)
## liqCode 0.000

# OA JuN
t.test(x = subset(data.intensity, liquid=="Juice" & group=="OA")$rating,
       y = subset(data.intensity, liquid=="Neutral" & group=="OA")$rating,
       paired = TRUE)

##
## Paired t-test
##
## data:  subset(data.intensity, liquid == "Juice" & group == "OA")$rating and subset(data.intensity, l
## t = 5.6293, df = 40, p-value = 1.568e-06
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  0.984905 2.088266
## sample estimates:
## mean of the differences
##                1.536585

# OA NuS
t.test(x = subset(data.intensity, liquid=="Saltwater" & group=="OA")$rating,
       y = subset(data.intensity, liquid=="Neutral" & group=="OA")$rating,
       paired = TRUE)

##
## Paired t-test
##
## data:  subset(data.intensity, liquid == "Saltwater" & group == "OA")$rating and subset(data.intensity,
## t = 1.8891, df = 40, p-value = 0.06615
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.05112942 1.51454405
## sample estimates:
## mean of the differences
##                0.7317073

# YA
m.selfreport.2.YA<-lmer(formula = rating ~ liqCode + (1|subID),
                      data = subset(data.intensity, group=="YA"))
summary(m.selfreport.2.YA)

```

```
## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode + (1 | subID)
## Data: subset(data.intensity, group == "YA")
##
## REML criterion at convergence: 648.9
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.0513 -0.8841  0.2285  0.7893  1.3865
##
## Random effects:
##   Groups   Name                Variance Std.Dev.
##   subID    (Intercept)  0.000      0.000
##   Residual                    3.179      1.783
## Number of obs: 162, groups:  subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   4.59259    0.14009 160.00000  32.784   <2e-16 ***
## liqCode       -0.06481    0.17157 160.00000  -0.378    0.706
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##          (Intr)
## liqCode 0.000
```

```
# YA JvN
```

```
t.test(x = subset(data.intensity, liquid=="Juice" & group=="YA")$rating,
       y = subset(data.intensity, liquid=="Neutral" & group=="YA")$rating,
       paired = TRUE)
```

```
##
## Paired t-test
##
## data:  subset(data.intensity, liquid == "Juice" & group == "YA")$rating and subset(data.intensity, l
## t = 8.2771, df = 53, p-value = 4.048e-11
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  1.571477 2.576671
## sample estimates:
## mean of the differences
##                2.074074
```

```
# YA NvS
```

```
t.test(x = subset(data.intensity, liquid=="Saltwater" & group=="YA")$rating,
       y = subset(data.intensity, liquid=="Neutral" & group=="YA")$rating,
       paired = TRUE)
```

```
##
## Paired t-test
##
## data:  subset(data.intensity, liquid == "Saltwater" & group == "YA")$rating and subset(data.intensit
## t = 8.4047, df = 53, p-value = 2.54e-11
## alternative hypothesis: true difference in means is not equal to 0
```

```

## 95 percent confidence interval:
## 1.677797 2.729610
## sample estimates:
## mean of the differences
## 2.203704

# OA vs YA
# Juice
t.test(x = subset(data.intensity, liquid=="Juice" & group=="OA")$rating,
       y = subset(data.intensity, liquid=="Juice" & group=="YA")$rating)

##
## Welch Two Sample t-test
##
## data: subset(data.intensity, liquid == "Juice" & group == "OA")$rating and subset(data.intensity, l
## t = 0.17111, df = 92.49, p-value = 0.8645
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.4065377 0.4831957
## sample estimates:
## mean of x mean of y
## 5.279070 5.240741

# Neutral
t.test(x = subset(data.intensity, liquid=="Neutral" & group=="OA")$rating,
       y = subset(data.intensity, liquid=="Neutral" & group=="YA")$rating)

##
## Welch Two Sample t-test
##
## data: subset(data.intensity, liquid == "Neutral" & group == "OA")$rating and subset(data.intensity,
## t = 1.8983, df = 77.973, p-value = 0.06136
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.02993791 1.25758018
## sample estimates:
## mean of x mean of y
## 3.780488 3.166667

# Saltwater
t.test(x = subset(data.intensity, liquid=="Saltwater" & group=="OA")$rating,
       y = subset(data.intensity, liquid=="Saltwater" & group=="YA")$rating)

##
## Welch Two Sample t-test
##
## data: subset(data.intensity, liquid == "Saltwater" & group == "OA")$rating and subset(data.intensit
## t = -2.7841, df = 94.138, p-value = 0.006488
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.5109895 -0.2530071
## sample estimates:
## mean of x mean of y
## 4.488372 5.370370

```

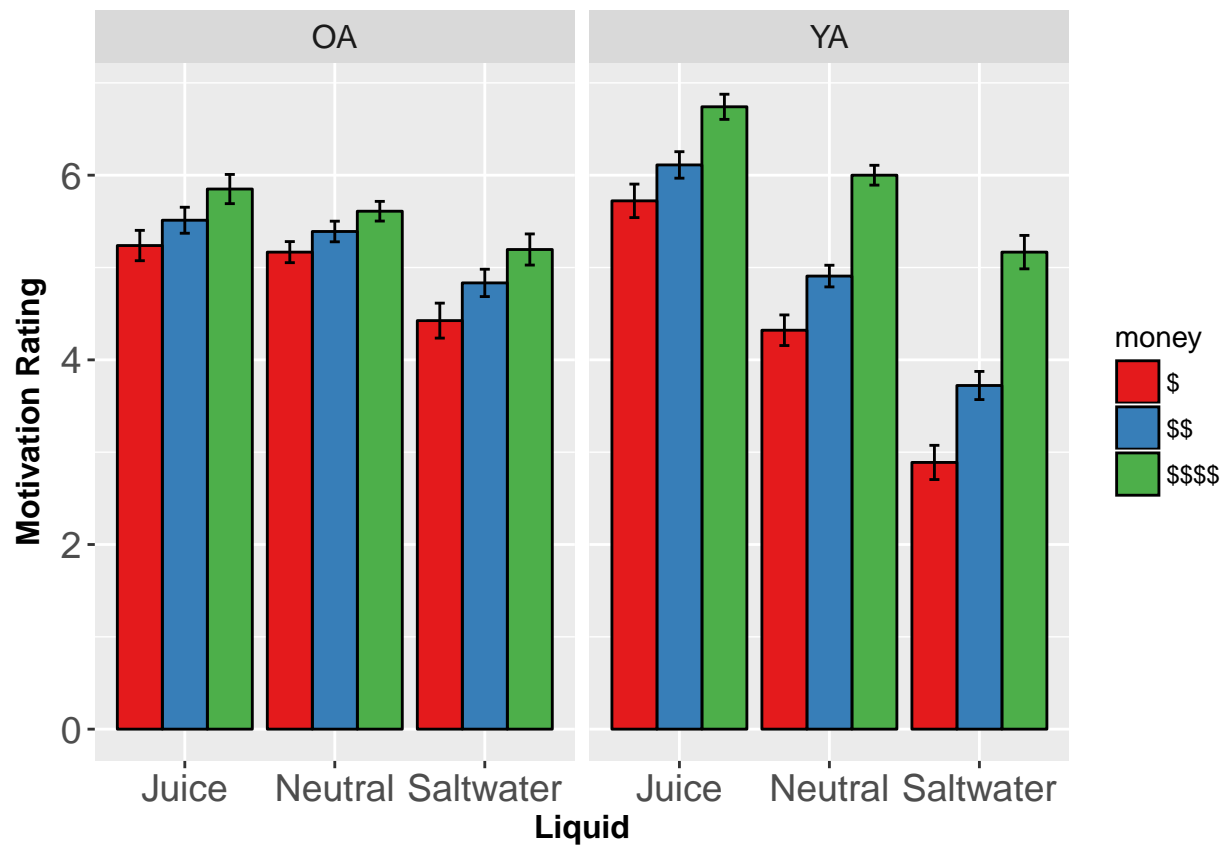
## Plot of Self-Report Motivation Ratings per Trial Type

```
# 9 motivation ratings
data.motive9<-data.selfreport %>%
  select(subID, group,
         Juice_1=juice1_motivation, Juice_2=juice2_motivation, Juice_4=juice4_motivation,
         Neutral_1=neut1_motivation, Neutral_2=neut2_motivation, Neutral_4=neut4_motivation,
         Saltwater_1=salt1_motivation, Saltwater_2=salt2_motivation, Saltwater_4=salt4_motivation) %>%
  gather(key = condition, value = rating, Juice_1:Saltwater_4, factor_key = TRUE) %>%
  separate(col = condition, into=c("liquid","moneyreward"), sep = "_") %>%
  mutate(liqCode = factor(liquid, levels = c("Saltwater","Neutral","Juice"), labels=c(-1,0,1)),
         rewCode = factor(moneyreward, levels = c(1,2,4), labels=c(-1,0,1)),
         groupCode = factor(group, levels=c("YA","OA"), labels=c(0,1)),
         money=factor(moneyreward, levels=c(1,2,4), labels=c("$","$$","$$$")))
data.motive9$liqCode<-as.numeric(levels(data.motive9$liqCode)[data.motive9$liqCode])
data.motive9$rewCode<-as.numeric(levels(data.motive9$rewCode)[data.motive9$rewCode])
data.motive9$groupCode<-as.numeric(levels(data.motive9$groupCode)[data.motive9$groupCode])
data.motive9$subID<-as.factor(data.motive9$subID)

selfreport.sum<-summarySEwithin2(data = data.motive9, measurevar = "rating",
                                withinvars = c("liquid","money"), betweenvars = "group",
                                idvar = "subID")

## Automatically converting the following non-factors to factors: liquid
p.rating.3<-ggplot(data = selfreport.sum, aes(x = liquid, y = rating, fill = money)) +
  geom_bar(position=position_dodge(), stat = "identity", color="black") +
  geom_errorbar(position=position_dodge(.9),
               mapping = aes(ymin=rating-se, ymax=rating+se), width=.2) +
  xlab("Liquid") + ylab("Motivation Rating") +
  scale_x_discrete(labels=c("juice_intense" = "Juice",
                           "neut_intense" = "Neutral",
                           "salt_intense" = "Saltwater")) +

  facet_grid(.~group) +
  scale_fill_brewer(palette = "Set1") +
  #coord_cartesian(ylim=c(1,7)) +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position="right",
        strip.text.x = element_text(size = 12))
p.rating.3
```



## Stats on Motivation Ratings by Trial Type

```
m.selfreport.3<-lmer(formula = rating ~ liqCode*rewCode*groupCode + (1|subID),
                     data = data.motive9)
summary(m.selfreport.3)
```

```
## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode * groupCode + (1 | subID)
## Data: data.motive9
##
## REML criterion at convergence: 2758.1
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4445 -0.5177  0.0441  0.5554  2.7243
##
## Random effects:
## Groups Name Variance Std.Dev.
## subID (Intercept) 1.134 1.065
## Residual 1.115 1.056
## Number of obs: 855, groups: subID, 96
##
## Fixed effects:
##
## Estimate Std. Error df t value Pr(>|t|)
## (Intercept) 5.06488 0.15262 93.90000 33.186 < 2e-16
## liqCode 1.13272 0.05867 753.00000 19.306 < 2e-16
## rewCode 0.82861 0.05877 753.10000 14.099 < 2e-16
## groupCode 0.18003 0.23089 94.10000 0.780 0.4375
## liqCode:rewCode -0.31481 0.07186 753.00000 -4.381 1.35e-05
## liqCode:groupCode -0.78471 0.08934 753.10000 -8.784 < 2e-16
## rewCode:groupCode -0.51760 0.08943 753.30000 -5.788 1.05e-08
## liqCode:rewCode:groupCode 0.26366 0.10974 753.60000 2.403 0.0165
##
## (Intercept) ***
## liqCode ***
## rewCode ***
## groupCode
## liqCode:rewCode ***
## liqCode:groupCode ***
## rewCode:groupCode ***
## liqCode:rewCode:groupCode *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) liqCod rewCod gropCd lqCd:rC lqCd:gC rwCd:C
## liqCode 0.000
## rewCode -0.001 0.000
## groupCode -0.661 0.000 0.001
## liqCod:rwCd 0.000 0.000 0.000 0.000
## liqCd:grpCd 0.000 -0.657 0.000 0.000 0.000
## rewCd:grpCd 0.001 0.000 -0.657 0.001 0.000 0.007
## lqCd:rwCd:C 0.000 0.000 0.000 0.003 -0.655 0.003 -0.003
```

```

# OA
m.selfreport.3.OA<-lmer(formula = rating ~ liqCode*rewCode + (1|subID),
                        data = subset(data.motive9, group=="OA"))
summary(m.selfreport.3.OA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode + (1 | subID)
## Data: subset(data.motive9, group == "OA")
##
## REML criterion at convergence: 1130
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.8790 -0.4159  0.0497  0.4907  2.7667
##
## Random effects:
##   Groups   Name                Variance Std.Dev.
##   subID    (Intercept)  1.4633     1.210
##   Residual                    0.8893     0.943
## Number of obs: 370, groups:  subID, 42
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    5.24479    0.19301  41.00000   27.174 < 2e-16 ***
## liqCode         0.34761    0.06016  325.10000    5.778 1.77e-08 ***
## rewCode         0.31137    0.06019  325.10000    5.173 4.03e-07 ***
## liqCode:rewCode -0.05172    0.07407  325.30000   -0.698  0.485
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod rewCod
## liqCode       0.000
## rewCode       0.002  0.012
## liqCod:rwCd   0.004  0.006 -0.006

# OA JvN
m.selfreport.3.OA.JvN<-lmer(formula = rating ~ liqCode*rewCode + (1|subID),
                           data = subset(data.motive9, group=="OA" & liquid!="Saltwater"))
summary(m.selfreport.3.OA.JvN)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode + (1 | subID)
## Data: subset(data.motive9, group == "OA" & liquid != "Saltwater")
##
## REML criterion at convergence: 686
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.1450 -0.3963  0.0026  0.4432  3.2916
##
## Random effects:
##   Groups   Name                Variance Std.Dev.

```



```

## subID (Intercept) 1.3809 1.1751
## Residual 0.5781 0.7604
## Number of obs: 247, groups: subID, 42
##
## Fixed effects:
## Estimate Std. Error df t value Pr(>|t|)
## (Intercept) 5.38965 0.19379 46.62000 27.811 <2e-16 ***
## liqCode 0.13017 0.09696 202.19000 1.343 0.1809
## rewCode 0.21684 0.08356 202.13000 2.595 0.0101 *
## liqCode:rewCode 0.08443 0.11861 202.18000 0.712 0.4774
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) liqCod rewCod
## liqCode -0.249
## rewCode 0.004 -0.008
## liqCod:rwCd -0.003 0.018 -0.705

# OA NvS
m.selfreport.3.OA.NvS<-lmer(formula = rating ~ liqCode*rewCode + (1|subID),
                           data = subset(data.motive9, group=="OA" & liquid!="Juice"))
summary(m.selfreport.3.OA.NvS)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode + (1 | subID)
## Data: subset(data.motive9, group == "OA" & liquid != "Juice")
##
## REML criterion at convergence: 734.4
##
## Scaled residuals:
## Min 1Q Median 3Q Max
## -4.1026 -0.4486 -0.0073 0.4802 2.8243
##
## Random effects:
## Groups Name Variance Std.Dev.
## subID (Intercept) 1.8472 1.3591
## Residual 0.6937 0.8329
## Number of obs: 247, groups: subID, 42
##
## Fixed effects:
## Estimate Std. Error df t value Pr(>|t|)
## (Intercept) 5.37799 0.22269 46.03000 24.150 < 2e-16 ***
## liqCode 0.54929 0.10620 202.14000 5.172 5.55e-07 ***
## rewCode 0.22138 0.09153 202.09000 2.419 0.0165 *
## liqCode:rewCode -0.19108 0.13038 202.16000 -1.466 0.1443
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) liqCod rewCod
## liqCode 0.237
## rewCode 0.004 0.008
## liqCod:rwCd 0.003 0.000 0.702

```

```

# YA
m.selfreport.3.YA<-lmer(formula = rating ~ liqCode*rewCode + (1|subID),
                        data = subset(data.motive9, group=="YA"))
summary(m.selfreport.3.YA)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode + (1 | subID)
## Data: subset(data.motive9, group == "YA")
##
## REML criterion at convergence: 1613.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.86104 -0.60892  0.02475  0.62075  2.54297
##
## Random effects:
##   Groups   Name              Variance Std.Dev.
## subID     (Intercept) 0.8799   0.938
## Residual                  1.2869   1.134
## Number of obs: 485, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    5.06485    0.13766  53.00000   36.794 < 2e-16 ***
## liqCode        1.13272    0.06302  428.00000   17.973 < 2e-16 ***
## rewCode        0.82865    0.06313  428.10000   13.126 < 2e-16 ***
## liqCode:rewCode -0.31481    0.07719  428.00000   -4.079 5.4e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) liqCod rewCod
## liqCode      0.000
## rewCode     -0.001  0.000
## liqCod:rwCd  0.000  0.000  0.000

# YA JvN
m.selfreport.3.YA.JvN<-lmer(formula = rating ~ liqCode*rewCode + (1|subID),
                            data = subset(data.motive9, group=="YA" & liquid!="Saltwater"))
summary(m.selfreport.3.YA.JvN)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode + (1 | subID)
## Data: subset(data.motive9, group == "YA" & liquid != "Saltwater")
##
## REML criterion at convergence: 985.5
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.2893 -0.5954  0.0439  0.6285  2.3025
##
## Random effects:
##   Groups   Name              Variance Std.Dev.

```

```

## subID (Intercept) 0.5075 0.7124
## Residual 0.9536 0.9765
## Number of obs: 323, groups: subID, 54
##
## Fixed effects:
## Estimate Std. Error df t value Pr(>|t|)
## (Intercept) 5.07501 0.12380 80.76000 40.992 <2e-16 ***
## liqCode 1.11635 0.10869 266.17000 10.270 <2e-16 ***
## rewCode 0.84119 0.09447 266.34000 8.904 <2e-16 ***
## liqCode:rewCode -0.33193 0.13324 266.22000 -2.491 0.0133 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) liqCod rewCod
## liqCode -0.441
## rewCode -0.005 0.006
## liqCod:rwCd 0.004 -0.004 -0.709

# YA NvS
m.selfreport.3.YA.NvS<-lmer(formula = rating ~ liqCode*rewCode + (1|subID),
                           data = subset(data.motive9, group=="YA" & liquid!="Juice"))
summary(m.selfreport.3.YA.NvS)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ liqCode * rewCode + (1 | subID)
## Data: subset(data.motive9, group == "YA" & liquid != "Juice")
##
## REML criterion at convergence: 1081.9
##
## Scaled residuals:
## Min 1Q Median 3Q Max
## -2.55379 -0.61570 0.01909 0.63339 2.43058
##
## Random effects:
## Groups Name Variance Std.Dev.
## subID (Intercept) 1.714 1.309
## Residual 1.112 1.055
## Number of obs: 323, groups: subID, 54
##
## Fixed effects:
## Estimate Std. Error df t value Pr(>|t|)
## (Intercept) 5.0791 0.1966 63.8700 25.830 < 2e-16 ***
## liqCode 1.1532 0.1174 266.0700 9.824 < 2e-16 ***
## rewCode 0.8350 0.1020 266.1400 8.184 1.15e-14 ***
## liqCode:rewCode -0.3038 0.1439 266.0900 -2.112 0.0357 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
## (Intr) liqCod rewCod
## liqCode 0.300
## rewCode -0.004 -0.006
## liqCod:rwCd -0.003 -0.004 0.709

```

```

# OA vs YA
# Juice
m.selfreport.3.OAYA.Juice<-lmer(formula = rating ~ groupCode*rewCode + (1|subID),
                                data = subset(data.motive9, liquid=="Juice"))
summary(m.selfreport.3.OAYA.Juice)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ groupCode * rewCode + (1 | subID)
## Data: subset(data.motive9, liquid == "Juice")
##
## REML criterion at convergence: 738.1
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.8615 -0.5779  0.0971  0.5903  2.7534
##
## Random effects:
##   Groups   Name      Variance Std.Dev.
## subID      (Intercept) 0.9830   0.9914
## Residual                0.3623   0.6019
## Number of obs: 285, groups: subID, 96
##
## Fixed effects:
##              Estimate Std. Error    df t value Pr(>|t|)
## (Intercept)    6.19136    0.14297  93.70000  43.307 < 2e-16 ***
## groupCode     -0.67007    0.21637  94.08000  -3.097  0.00258 **
## rewCode        0.50926    0.05792 187.01000   8.793 8.88e-16 ***
## groupCode:rewCode -0.20633    0.08842 187.38000  -2.333 0.02069 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) gropCd rewCod
## groupCode   -0.661
## rewCode      0.000  0.000
## gropCd:rwCd  0.000  0.005 -0.655

# Neutral
m.selfreport.3.OAYA.Neutral<-lmer(formula = rating ~ groupCode*rewCode + (1|subID),
                                   data = subset(data.motive9, liquid=="Neutral"))
summary(m.selfreport.3.OAYA.Neutral)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ groupCode * rewCode + (1 | subID)
## Data: subset(data.motive9, liquid == "Neutral")
##
## REML criterion at convergence: 843.4
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -2.68395 -0.46180 -0.03077  0.40484  2.72582
##
## Random effects:

```

```

## Groups      Name      Variance Std.Dev.
## subID      (Intercept) 1.496    1.223
## Residual                0.517    0.719
## Number of obs: 285, groups: subID, 96
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)      5.07555    0.17584  94.10000  28.864 < 2e-16 ***
## groupCode         0.31211    0.26596  94.25000   1.174  0.244
## rewCode           0.84038    0.06965 187.47000  12.066 < 2e-16 ***
## groupCode:rewCode -0.62687    0.10542 187.51000  -5.947 1.31e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) gropCd rewCod
## groupCode   -0.661
## rewCode      -0.003  0.002
## gropCd:rwCd  0.002  0.001 -0.661

# Saltwater
m.selfreport.3.OAYA.Saltwater<-lmer(formula = rating ~ groupCode*rewCode + (1|subID),
                                   data = subset(data.motive9, liquid=="Saltwater"))
summary(m.selfreport.3.OAYA.Saltwater)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: rating ~ groupCode * rewCode + (1 | subID)
## Data: subset(data.motive9, liquid == "Saltwater")
##
## REML criterion at convergence: 974.1
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.2309 -0.5394  0.0077  0.4607  2.4381
##
## Random effects:
## Groups      Name      Variance Std.Dev.
## subID      (Intercept) 2.6002    1.6125
## Residual                0.7925    0.8902
## Number of obs: 285, groups: subID, 96
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)      3.92593    0.23031  93.78000  17.046 < 2e-16 ***
## groupCode         0.90804    0.34851  94.10000   2.606  0.0107 *
## rewCode           1.13889    0.08566 187.04000  13.295 < 2e-16 ***
## groupCode:rewCode -0.72787    0.13146 187.55000  -5.537 1.03e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) gropCd rewCod
## groupCode   -0.661
## rewCode      0.000  0.000

```

```
## gropCd:rwCd  0.000 -0.002 -0.652
```

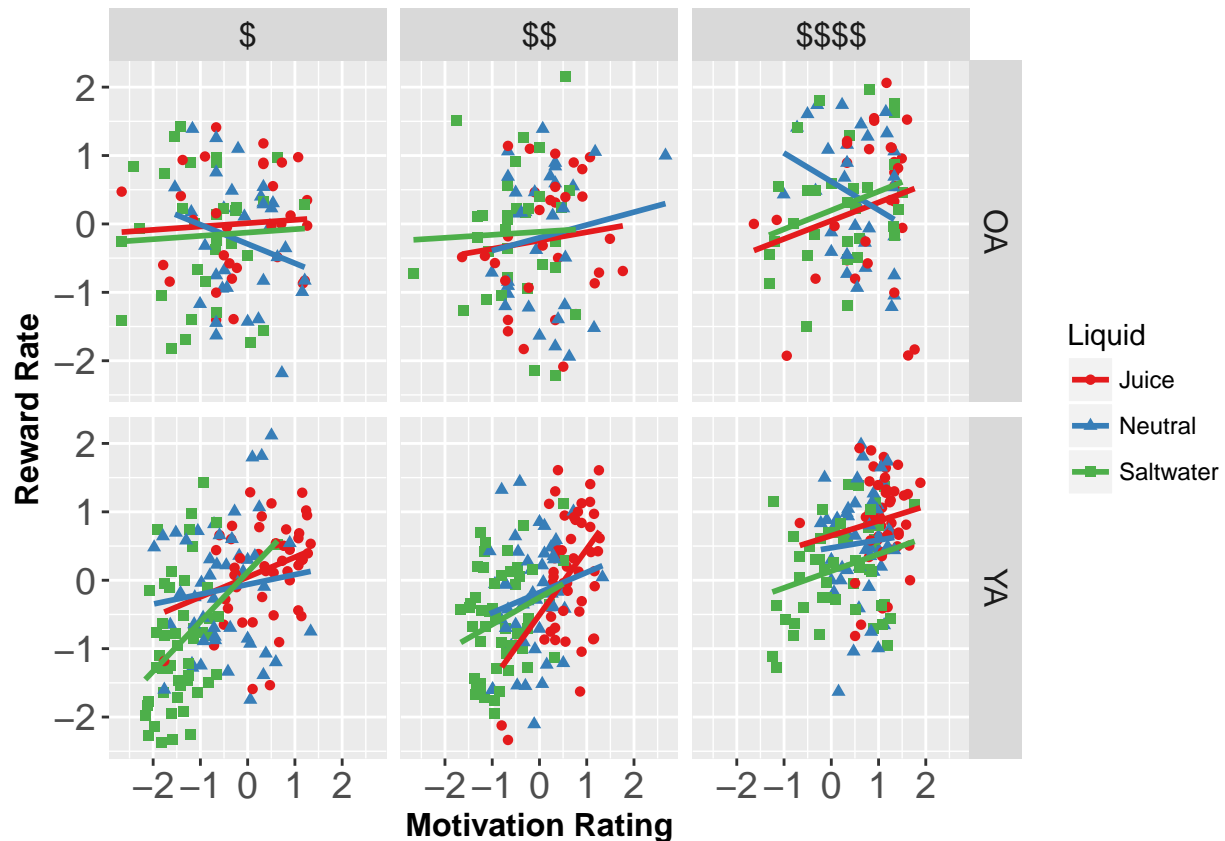
## Comparing Motivation Ratings and Reward Rate

```
incentive9.means.indiv <- incentive9.means %>%
  mutate(liquid = factor(liquid, levels = c("saltwater", "neutral", "juice"),
                        labels=c("Saltwater", "Neutral", "Juice"))) %>%
  inner_join(y = data.motive9, by = c("subID", "liquid", "money", "group")) %>%
  group_by(subID) %>%
  mutate(rating_zscore = scale(rating),
         meanRR_zscore = scale(meanRR))

## Warning: Column `liquid` joining factor and character vector, coercing into
## character vector

p.rating.4<-ggplot(data = incentive9.means.indiv,
                  aes(x = rating_zscore, y = meanRR_zscore, color=liquid, shape=liquid)) +
  geom_point() +
  geom_smooth(method="lm", se = FALSE) +
  xlab("Motivation Rating") + ylab("Reward Rate") +
  labs(shape = "Liquid", color="Liquid") +
  scale_color_brewer(palette="Set1") +
  facet_grid(group~money) +
  theme(#plot.title=element_text(size=22,face="bold", vjust=2),
        axis.title=element_text(size=12,face = "bold"),
        axis.text=element_text(size=14),
        legend.position="right",
        strip.text.x = element_text(size = 12),
        strip.text.y = element_text(size = 12))
p.rating.4

## Warning: Removed 148 rows containing non-finite values (stat_smooth).
## Warning: Removed 148 rows containing missing values (geom_point).
```



```
incentive9.means.indiv2<-filter(incentive9.means.indiv, is.na(rating)!=TRUE)
# OA Hierarchical Regression
# Model 1
m.selfreport.4.OA.L1<-lmer(formula = meanRR ~ rewCode*liqCode +
                           (1+rewCode|subID),
                           data = subset(incentive9.means.indiv2, group=="OA"))
summary(m.selfreport.4.OA.L1)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: meanRR ~ rewCode * liqCode + (1 + rewCode | subID)
## Data: subset(incentive9.means.indiv2, group == "OA")
##
## REML criterion at convergence: -444.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.4096 -0.6091  0.0769  0.6481  2.3120
##
## Random effects:
##  Groups   Name                Variance Std.Dev. Corr
##  subID    (Intercept)  0.0119168  0.10916
##           rewCode      0.0001556  0.01247  -1.00
##  Residual                    0.0127400  0.11287
## Number of obs: 370, groups:  subID, 42
##
## Fixed effects:
```



```

##               Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    0.573834   0.017841  41.000000   32.164  <2e-16 ***
## rewCode        0.021475   0.007456 166.000000    2.880   0.0045 **
## liqCode        0.009036   0.007200 325.000000    1.255   0.2104
## rewCode:liqCode -0.003049   0.008863 325.400000   -0.344   0.7310
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr) rewCod liqCod
## rewCode    -0.242
## liqCode      0.000  0.012
## rewCod:liqCd  0.005 -0.006  0.005

# Model 2
m.selfreport.4.OA.L2<-lmer(formula = meanRR ~ rewCode*liqCode + rating +
                           (1+rewCode|subID),
                           data = subset(incentive9.means.indiv2, group=="OA"))
summary(m.selfreport.4.OA.L2)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: meanRR ~ rewCode * liqCode + rating + (1 + rewCode | subID)
## Data: subset(incentive9.means.indiv2, group == "OA")
##
## REML criterion at convergence: -441.4
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.3915 -0.5970  0.0860  0.6394  2.2731
##
## Random effects:
##   Groups   Name      Variance Std.Dev. Corr
##   subID    (Intercept) 0.0117185 0.10825
##           rewCode     0.0001474 0.01214  -1.00
## Residual                0.0126074 0.11228
## Number of obs: 370, groups: subID, 42
##
## Fixed effects:
##               Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)    0.503239   0.035935 213.400000   14.004  <2e-16 ***
## rewCode        0.017293   0.007634 182.200000    2.265   0.0247 *
## liqCode        0.004331   0.007459 332.300000    0.581   0.5619
## rating         0.013461   0.005963 339.700000    2.257   0.0246 *
## rewCode:liqCode -0.002379   0.008822 324.900000   -0.270   0.7876
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##           (Intr) rewCod liqCod rating
## rewCode      0.098
## liqCode       0.243  0.079
## rating       -0.870 -0.242 -0.279
## rewCod:liqCd -0.027 -0.014 -0.004  0.034

```

```

# Model Comparison
anova(m.selfreport.4.OA.L1,m.selfreport.4.OA.L2)

## refitting model(s) with ML (instead of REML)

## Data: subset(incentive9.means.indiv2, group == "OA")
## Models:
## object: meanRR ~ rewCode * liqCode + (1 + rewCode | subID)
## ..1: meanRR ~ rewCode * liqCode + rating + (1 + rewCode | subID)
##      Df      AIC      BIC logLik deviance Chisq Chi Df Pr(>Chisq)
## object  8 -458.63 -427.32 237.31  -474.63
## ..1     9 -461.76 -426.54 239.88  -479.76 5.1321      1 0.02349 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# YA Hierarchical Regression
# Model 1
m.selfreport.4.YA.L1<-lmer(formula = meanRR ~ rewCode*liqCode +
                           (1+rewCode|subID),
                           data = subset(incentive9.means.indiv2, group=="YA"))
summary(m.selfreport.4.YA.L1)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: meanRR ~ rewCode * liqCode + (1 + rewCode | subID)
## Data: subset(incentive9.means.indiv2, group == "YA")
##
## REML criterion at convergence: -514.7
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.6105 -0.5431  0.0581  0.5902  2.8398
##
## Random effects:
## Groups   Name                Variance Std.Dev. Corr
## subID    (Intercept)  1.173e-02 0.108312
##          rewCode      5.955e-05 0.007717 -1.00
## Residual                1.523e-02 0.123427
## Number of obs: 485, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept)   0.644515   0.015769  53.000000  40.871 < 2e-16 ***
## rewCode       0.053965   0.006949  316.200000   7.766 1.14e-13 ***
## liqCode       0.058546   0.006857  428.000000   8.538 2.22e-16 ***
## rewCode:liqCode -0.015480   0.008398  428.000000  -1.843  0.066 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) rewCod liqCod
## rewCode      -0.142
## liqCode       0.000  0.000
## rewCod:liqCd  0.000  0.000  0.000

```

```

# Model 2
m.selfreport.4.YA.L2<-lmer(formula = meanRR ~ rewCode*liqCode + rating +
                           (1+rewCode|subID),
                           data = subset(incentive9.means.indiv2, group=="YA"))
summary(m.selfreport.4.YA.L2)

## Linear mixed model fit by REML t-tests use Satterthwaite approximations
## to degrees of freedom [lmerMod]
## Formula: meanRR ~ rewCode * liqCode + rating + (1 + rewCode | subID)
## Data: subset(incentive9.means.indiv2, group == "YA")
##
## REML criterion at convergence: -587.5
##
## Scaled residuals:
##      Min       1Q   Median       3Q      Max
## -3.2698 -0.5695  0.0408  0.6759  2.6526
##
## Random effects:
## Groups Name Variance Std.Dev. Corr
## subID (Intercept) 0.011328 0.10643
##      rewCode 0.000128 0.01131 -1.00
## Residual 0.012690 0.11265
## Number of obs: 485, groups: subID, 54
##
## Fixed effects:
##              Estimate Std. Error      df t value Pr(>|t|)
## (Intercept) 0.423894 0.027894 287.900000 15.197 <2e-16 ***
## rewCode 0.017857 0.007496 296.800000 2.382 0.0178 *
## liqCode 0.009204 0.008141 455.000000 1.130 0.2589
## rating 0.043561 0.004597 477.800000 9.476 <2e-16 ***
## rewCode:liqCode -0.001767 0.007800 429.700000 -0.226 0.8209
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Correlation of Fixed Effects:
##              (Intr) rewCod liqCod rating
## rewCode 0.317
## liqCode 0.534 0.325
## rating -0.835 -0.508 -0.640
## rewCod:liqCd -0.155 -0.094 -0.119 0.186

# Model Comparison
anova(m.selfreport.4.YA.L1, m.selfreport.4.YA.L2)

## refitting model(s) with ML (instead of REML)
## Data: subset(incentive9.means.indiv2, group == "YA")
## Models:
## object: meanRR ~ rewCode * liqCode + (1 + rewCode | subID)
## ..1: meanRR ~ rewCode * liqCode + rating + (1 + rewCode | subID)
##      Df      AIC      BIC logLik deviance Chisq Chi Df Pr(>Chisq)
## object 8 -529.13 -495.65 272.56 -545.13
## ..1 9 -609.53 -571.87 313.76 -627.53 82.401 1 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

## Manuscript Figures

```
ms.path="/Users/debbieyee/Documents/Manuscripts/In Progress/Motivation Aging/Figures/"
ggsave(filename = "Figure2a_group.eps", plot = p.RR.0,
        device=cairo_ps, fallback_resolution = 600,
        path = ms.path, width = 3, height = 6, scale = 1)
ggsave(filename = "Figure2b_RT.eps", plot = p.RT.0, device = "eps",
        path = ms.path, width = 6, height = 6, scale = 1)
ggsave(filename = "Figure2c_ACC.eps", plot = p.ACC.0, device = "eps",
        path = ms.path, width = 6, height = 6, scale = 1)
ggsave(filename = "Figure3_Transition.eps", plot = p.trans.1,
        device = cairo_ps,
        path = ms.path, width = 12, height = 4, scale = 1)
ggsave(filename = "Figure4a_RR-mon-liq.eps", plot = p.RR.2,
        device = "eps",
        path = ms.path, width = 8, height = 4, scale = 1)
ggsave(filename = "Figure4b_RR-mon-block.eps", plot = p.RR.3,
        device = "eps",
        path = ms.path, width = 8, height = 4, scale = 1)
ggsave(filename = "Figure5_ValEst.eps", plot = p.val.1,
        device = cairo_ps,
        path = ms.path, width = 10, height = 6, scale = 1)
ggsave(filename = "Figure6a_Liking.eps", plot = p.rating.1,
        device = "eps",
        path = ms.path, width = 5.5, height = 5, scale = 1)
ggsave(filename = "Figure6b_Intensity.eps", plot = p.rating.2,
        device = "eps",
        path = ms.path, width = 5.5, height = 5, scale = 1)
ggsave(filename = "Figure6c_Motivation.eps", plot = p.rating.4,
        device = "eps",
        path = ms.path, width = 12, height = 6, scale = 1)
```

```
## Warning: Removed 148 rows containing non-finite values (stat_smooth).
```

```
## Warning: Removed 148 rows containing missing values (geom_point).
```