

Final 241 Project Consolidation

author: Elizabeth Khan, Estrella Ndrianasy, Chandni Shah, Michelle Shen, Catherine Tsai date: “November 8, 2021” output: html_document —

```
# Import data and remove invalid age

d <- fread("survey_cleaned_with_states.csv")

d <- d[Age!='Other']

# Removing company test response 52.36
d <- d[ResponseId!='R_2WC5v1B2DCvHHRH']

degrees = c("Bachelors degree", "Associates degree", "Masters degree")

# create column to indicate whether or not subject has a college degree

# d<- d[,college_degree:= ifelse(Education %in% degrees, 1,0)]

head(d)
```

```
##           ResponseId Gender   Age           Education LocationLatitude
## 1: R_x2RQnFD8LZIZuRH Female 40-50 Associates degree      36.99040
## 2: R_09f8Af9asil81DH Female 30-40 Some College No degree      27.99290
## 3: R_3oTqFe6v8l4jV0p Female 30-40 Some College No degree      36.83070
## 4: R_0ePUcdKMDE2jNe1 Female 30-40 Bachelors degree      39.30370
## 5: R_2owAeGSUxdN9fQR Female 40-50 Associates degree      44.14149
## 6: R_3MR7piXWpVNthOP Female 30-40 Some College No degree      38.88420
## LocationLongitude instr_enthusiasm instr_professional instr_subject
## 1:          -85.9239              4              4              4
## 2:          -82.4030              1              4              3
## 3:          -76.3146              2              2              2
## 4:          -94.9322              3              1              1
## 5:         -103.2052              4              4              5
## 6:          -76.9941              3              3              3
## instr_material instr_rating              subject_1
## 1:              4              4              Web Applications
## 2:              2              4              Web Applications
## 3:              2              2              Web Applications
## 4:              3              2 Artificial intelligence projects
## 5:              4              4              Web Applications
## 6:              3              3              All of the above
##              subject_2
## 1:              Both A & B
## 2: C. A low level implementation programming language
## 3:              B. A flexible programming language
```

```

## 4:          B. A flexible programming language
## 5:          B. A flexible programming language
## 6:          All of the above
##
## 1: Python is a compiled programming language with a faster and more efficient execution time than in
## 2:          Python is an object
## 3:          Python was crea
## 4:          Python was crea
## 5:          Python is an objec
## 6:          Python is an i
##
##          subject_4          video_watched assigned assignment ques_1
## 1: Great Advanced Language Video3-2(Control)      0    Control    0
## 2:          Easy to Learn Video2-2(Control)      0    Control    0
## 3:   Wonderful Community Video2-1(Treatment)      1  Treatment    0
## 4:   Wonderful Community Video1-2(Control)      0    Control    0
## 5:   Wonderful Community Video2-2(Control)      0    Control    0
## 6: Great Starter Language Video3-2(Control)      0    Control    1
##
##      ques_2 ques_3 ques_4 quiz_avg avg_rating          state state_code
## 1:      1      1      1    0.75    4.00      kentucky      KY
## 2:      0      0      0    0.00    2.75      florida      FL
## 3:      0      0      0    0.00    2.00      virginia      VA
## 4:      0      0      0    0.00    2.25      kansas      KS
## 5:      0      0      0    0.00    4.00      south dakota      SD
## 6:      0      0      0    0.25    3.00 district of columbia      DC
##
##      dem_percent rep_percent
## 1:      0.362      0.621
## 2:      0.479      0.512
## 3:      0.541      0.440
## 4:      0.416      0.562
## 5:      0.356      0.618
## 6:      0.921      0.054

```

Figures and Tables

```

summary_response_table <- d %>%
  count(Gender, Age, assignment)

```

```
summary_response_table
```

```

##      Gender  Age assignment  n
## 1: Female 20-30    Control 14
## 2: Female 20-30  Treatment 11
## 3: Female 30-40    Control 27
## 4: Female 30-40  Treatment 29
## 5: Female 40-50    Control 15
## 6: Female 40-50  Treatment 17
## 7: Female  50+    Control 12
## 8: Female  50+  Treatment 17
## 9:   Male 20-30    Control  9
## 10:  Male 20-30  Treatment 10
## 11:  Male 30-40    Control 12
## 12:  Male 30-40  Treatment 14

```

```
## 13:   Male 40-50   Control 11
## 14:   Male 40-50 Treatment  5
## 15:   Male  50+   Control  8
## 16:   Male  50+ Treatment 10
```

```
## Table printed with 'knitr::kable()', not {gt}. Learn why at
## http://www.danielsjoberg.com/gtsummary/articles/rmarkdown.html
## To suppress this message, include 'message = FALSE' in code chunk header.
```

Characteristic	Overall , N = 221	Female , N = 142	Male , N = 79
assignment			
Control	108 (49%)	68 (48%)	40 (51%)
Treatment	113 (51%)	74 (52%)	39 (49%)
Age			
20-30	44 (20%)	25 (18%)	19 (24%)
30-40	82 (37%)	56 (39%)	26 (33%)
40-50	48 (22%)	32 (23%)	16 (20%)
50+	47 (21%)	29 (20%)	18 (23%)

```
d %>%
  select(Gender, Education, assignment) %>%
  tbl_summary(by = assignment) %>%
  add_overall()
```

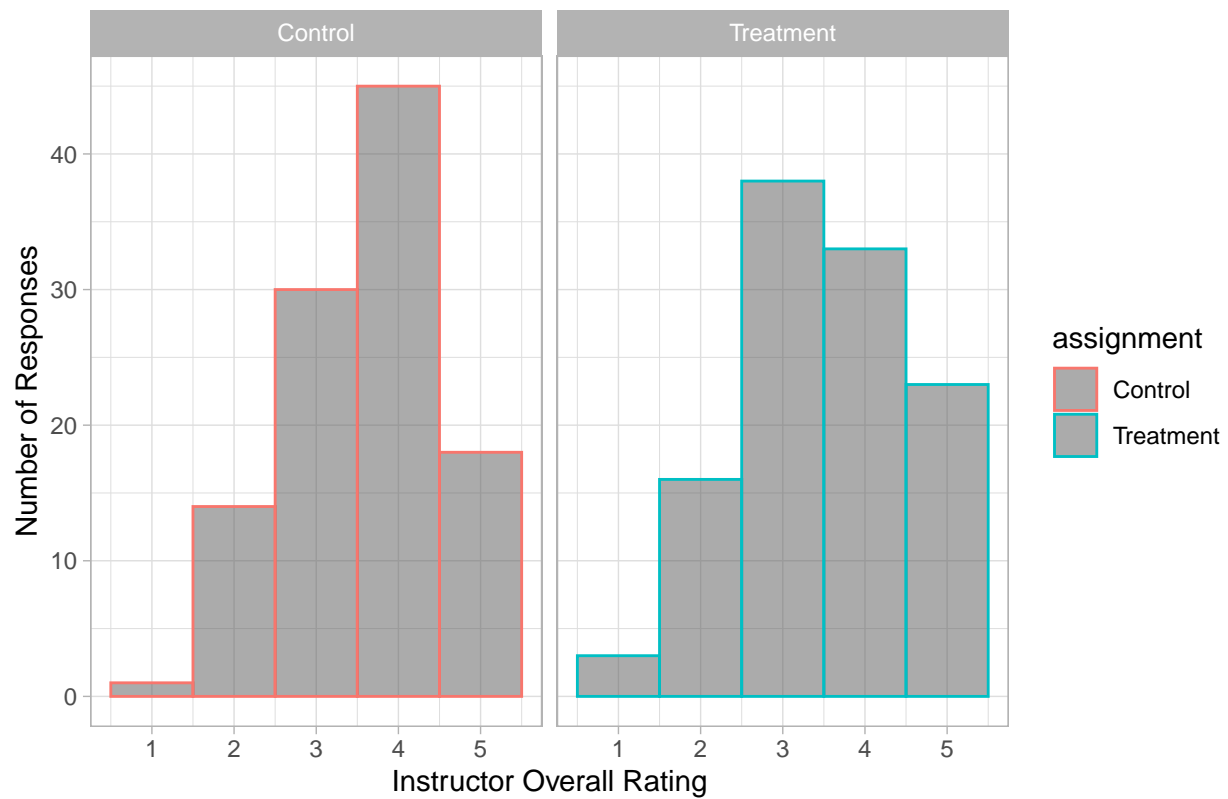
```
## Table printed with 'knitr::kable()', not {gt}. Learn why at
## http://www.danielsjoberg.com/gtsummary/articles/rmarkdown.html
## To suppress this message, include 'message = FALSE' in code chunk header.
```

Characteristic	Overall , N = 221	Control , N = 108	Treatment , N = 113
Gender			
Female	142 (64%)	68 (63%)	74 (65%)
Male	79 (36%)	40 (37%)	39 (35%)
Education			
Associates degree	27 (12%)	16 (15%)	11 (9.7%)
Bachelors degree	43 (19%)	21 (19%)	22 (19%)
High school diploma	66 (30%)	33 (31%)	33 (29%)
Less than High school	7 (3.2%)	1 (0.9%)	6 (5.3%)
Masters degree	11 (5.0%)	5 (4.6%)	6 (5.3%)
Some College No degree	67 (30%)	32 (30%)	35 (31%)

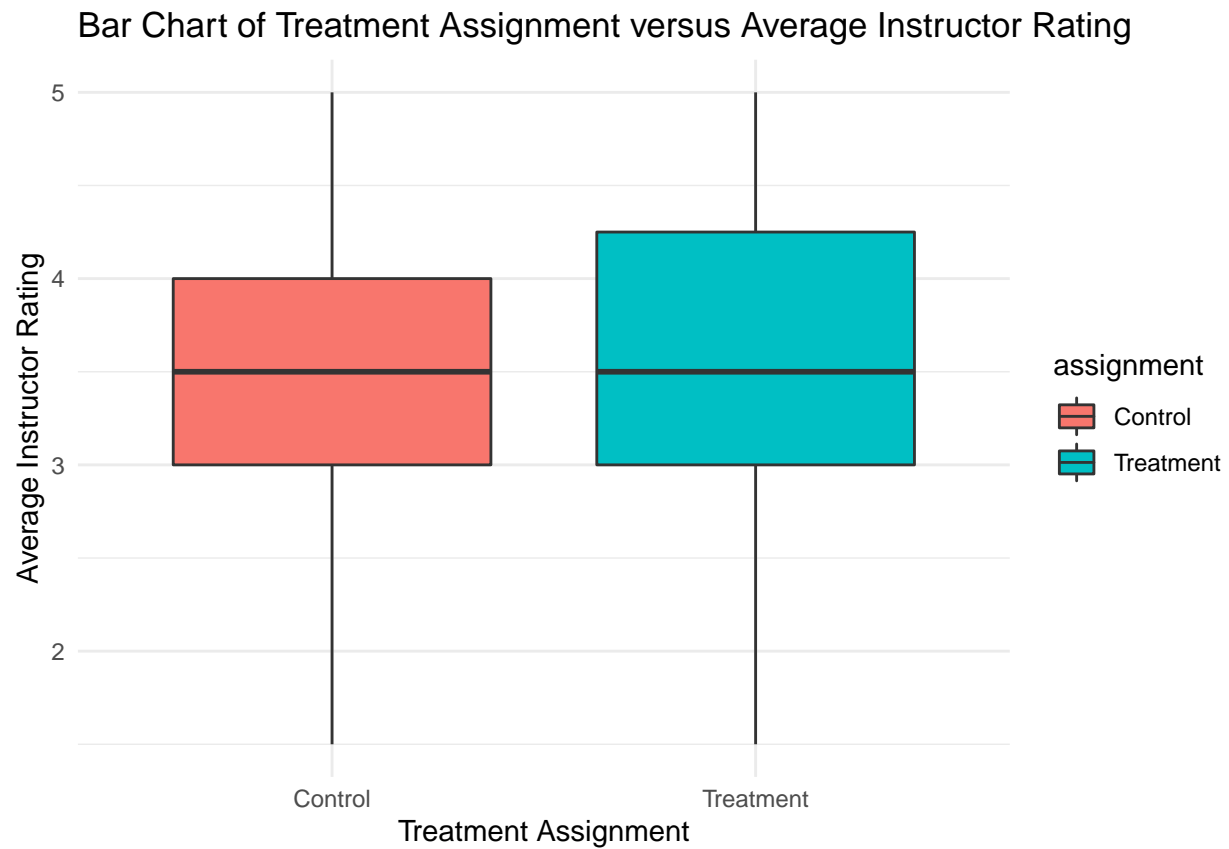
```
response_hist <- d %>%
  ggplot() +
  aes(x = instr_rating, color=assignment) +
  geom_histogram(alpha = 0.5, bins=5) +
  labs(
    title = 'Instructor Overall Rating by Treatment and Control',
    x     = 'Instructor Overall Rating',
    y     = 'Number of Responses'
  ) +
  theme_light() +
  facet_wrap(~ assignment)

response_hist
```

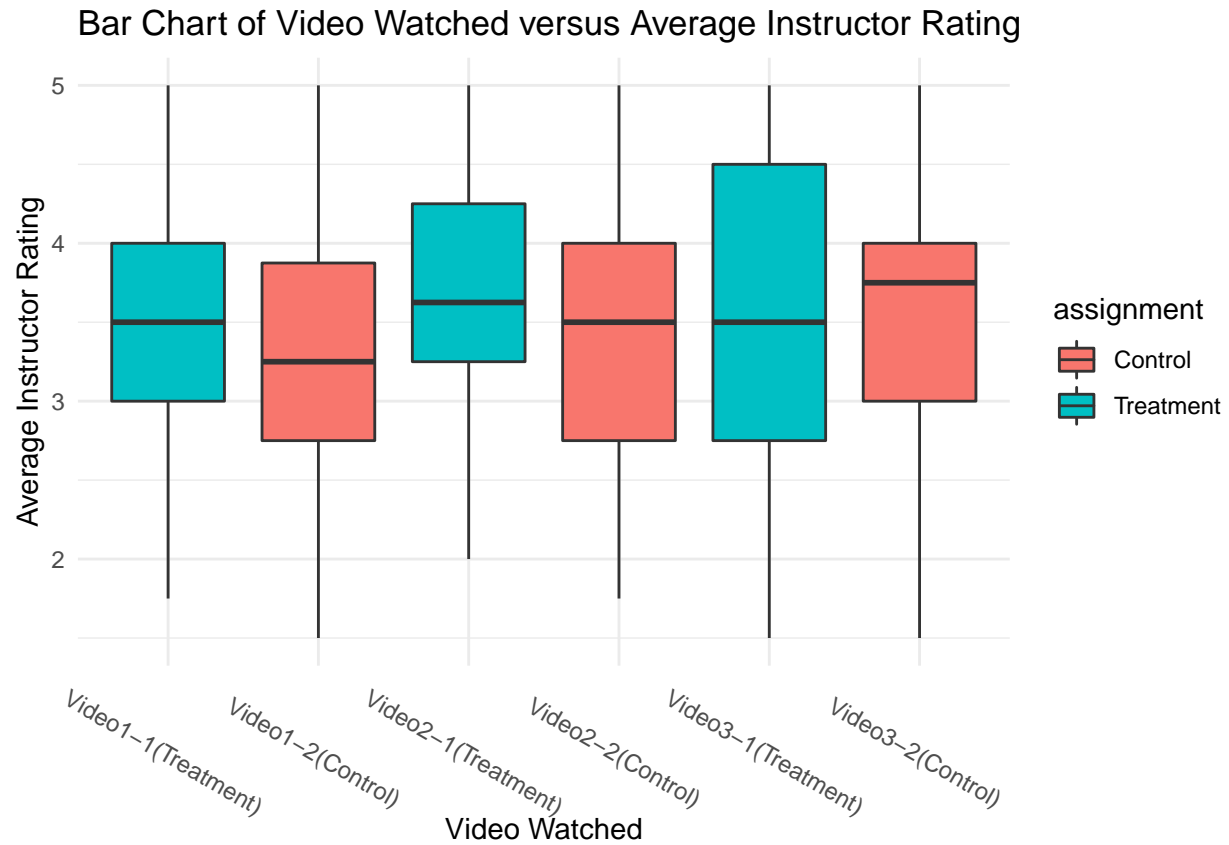
Instructor Overall Rating by Treatment and Control



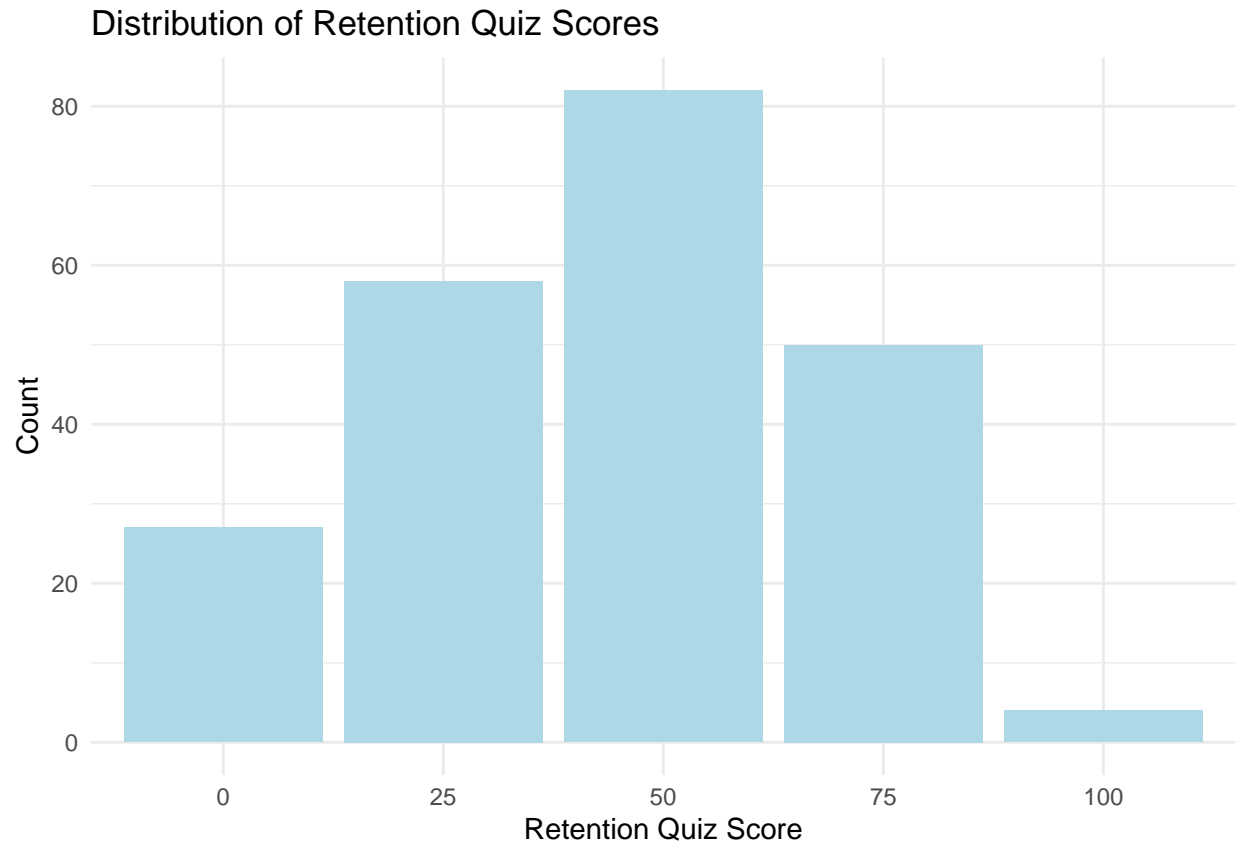
```
ggplot(d, aes(x=assignment, y=avg_rating, fill=assignment))+
  geom_boxplot()+
  labs(title="Bar Chart of Treatment Assignment versus Average Instructor Rating",
        x ="Treatment Assignment", y = "Average Instructor Rating")
```



```
ggplot(d, aes(x=video_watched, y=avg_rating, fill=assignment))+  
  geom_boxplot()+  
  labs(title="Bar Chart of Video Watched versus Average Instructor Rating",  
        x="Video Watched", y="Average Instructor Rating")+theme(axis.text.x = element_text(angle = -45))
```



```
d$quiz_grade <- as.factor(d$quiz_avg*100)
ggplot(d, aes(x=quiz_grade))+
  geom_bar(aes(y = (...count...)),fill="lightblue")+
  labs(title="Distribution of Retention Quiz Scores",
        x="Retention Quiz Score", y="Count")
```



Randomness Check

In this section, we verify if the block randomization was successful; there should be no increase in predicting whether or not someone is in the treatment or control group based on their blocks.

```
null_model <- lm(assigned~1,data=d)
full_model <- lm(assigned ~ 1 + as.factor(Age)+as.factor(Gender), data=d)
f_test <- anova(full_model, null_model, test = 'F')
f_test
```

```
## Analysis of Variance Table
##
## Model 1: assigned ~ 1 + as.factor(Age) + as.factor(Gender)
## Model 2: assigned ~ 1
##   Res.Df    RSS Df Sum of Sq    F Pr(>F)
## 1     216 54.797
## 2     220 55.222 -4   -0.42483 0.4186 0.7951
```

```
covs <- c('Age','Gender','Education', 'assignment','avg_rating')
subset <- d[,..covs]

head(subset)
```

```
##      Age Gender      Education assignment avg_rating
```

```
options(qwraps2_markup = 'markdown')
covariate_summary <-
  list("Gender" =
    list("Male"      = ~ qwraps2::n_perc(Gender == 'Male'),
         "Female"    = ~ qwraps2::n_perc(Gender == 'Female')),
    "Age" =
    list("20-30" = ~ qwraps2::n_perc(Age == '20-30'),
         "30-40" = ~ qwraps2::n_perc(Age == '30-40'),
         "40-50" = ~ qwraps2::n_perc(Age == '40-50'),
         "50+"  = ~ qwraps2::n_perc(Age == '50+')),
    "Education" =
    list("Less than High school" = ~ qwraps2::n_perc(Education == 'Less than High school'),
         "High school diploma"   = ~ qwraps2::n_perc(Education == 'High school diploma'),
         "Some College No degree" = ~ qwraps2::n_perc(Education == 'Some College No degree'),
         "Associates degree"      = ~ qwraps2::n_perc(Education == 'Associates degree'),
         "Bachelors degree"       = ~ qwraps2::n_perc(Education == 'Bachelors degree'),
         "Masters degree"         = ~ qwraps2::n_perc(Education == 'Masters degree'))
  )

table <- summary_table(subset, covariate_summary, by = c("assignment"))

table
```

	Control (N = 108)	Treatment (N = 113)
:	:-----	:-----
Gender		
Male	40 (37.04%)	39 (34.51%)
Female	68 (62.96%)	74 (65.49%)
Age		
20-30	23 (21.30%)	21 (18.58%)
30-40	39 (36.11%)	43 (38.05%)
40-50	26 (24.07%)	22 (19.47%)
50+	20 (18.52%)	27 (23.89%)
Education		
Less than High school	1 (0.93%)	6 (5.31%)
High school diploma	33 (30.56%)	33 (29.20%)
Some College No degree	32 (29.63%)	35 (30.97%)
Associates degree	16 (14.81%)	11 (9.73%)
Bachelors degree	21 (19.44%)	22 (19.47%)
Masters degree	5 (4.63%)	6 (5.31%)

```
## [1] "Associates degree"      "Some College No degree" "Bachelors degree"
```



```
## [4] "Less than High school" "High school diploma" "Masters degree"
```

```
ages <- unique(d[,Age])
p_values_ages <- data.table(Age = character(), mean_rating_control = numeric(), mean_rating_treatment = numeric())

for(age in ages)
{
  print(age)

  treatment_avg = round(mean(d[(Age==age)&(assignment=='Treatment')],instr_rating),2)
  control_avg = round(mean(d[(Age==age)&(assignment=='Control')],instr_rating),2)
  p_val =round(t.test(d[(Age==age)&(assignment=='Treatment')],instr_rating), d[(Age==age)&(assignment=='Control')],instr_rating)$p.value,2)
  table1 <-data.table(Age=age,mean_rating_control = control_avg, mean_rating_treatment = treatment_avg, p_val=p_val)
  p_values_ages<-rbind(p_values_ages, table1)
}
```

```
## [1] "40-50"
## [1] "30-40"
## [1] "50+"
## [1] "20-30"
```

```
p_values_ages <- unique(p_values_ages)
```

```
t.test(d[(Age==age)&(assignment=='Treatment')],instr_rating), d[(Age==age)&(assignment=='Control')],instr_rating)
```

```
##
## Welch Two Sample t-test
##
## data: d[(Age == age) & (assignment == "Treatment"), instr_rating] and d[(Age == age) & (assignment == "Control"), instr_rating]
## t = 0.16838, df = 40.218, p-value = 0.8671
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.5921768 0.6998373
## sample estimates:
## mean of x mean of y
## 3.619048 3.565217
```

```
print(age)
```

```
## [1] "20-30"
```

```
genders <- unique(d[,Gender])
p_values_genders<- data.table(Gender = character(), mean_rating_control = numeric(), mean_rating_treatment = numeric())

for(gender in genders)
{
  print(gender)
```

```

treatment_avg =round(mean(d[(Gender==gender)&(assignment=='Treatment'),avg_rating]),2)
control_avg =round(mean(d[(Gender==gender)&(assignment=='Control'),avg_rating]),2)
p_val =round(t.test(d[(Gender==gender)&(assignment=='Treatment'),avg_rating], d[(Gender==gender)&(assignment=='Control'),avg_rating]),2)
table1 <-data.table(Gender=gender,mean_rating_control = control_avg, mean_rating_treatment = treatment_avg, p_value=p_val)
p_values_genders<-rbind(p_values_genders, table1)
}

```

```

## [1] "Female"
## [1] "Male"

```

```

educations <- unique(d[,Education])
p_values_educations<- data.table(Education = character(), mean_rating_control = numeric(), mean_rating_treatment = numeric(), p_value = numeric())

for(education in educations)

{ print(education)
treatment_avg =round(mean(d[(Education==education)&(assignment=='Treatment'),avg_rating]),2)
control_avg =round(mean(d[(Education==education)&(assignment=='Control'),avg_rating]),2)
if (education!= 'Less than High school') {
p_val =round(t.test(d[(Education==education)&(assignment=='Treatment'),avg_rating], d[(Education==education)&(assignment=='Control'),avg_rating]),2)
else {p_val = NA}
table1 <-data.table(Education=education,mean_rating_control = control_avg, mean_rating_treatment = treatment_avg, p_value=p_val)
p_values_educations<-rbind(p_values_educations, table1)
}

```

```

## [1] "Associates degree"
## [1] "Some College No degree"
## [1] "Bachelors degree"
## [1] "Less than High school"
## [1] "High school diploma"
## [1] "Masters degree"

```

```

p_values_all <- left_join(subset, p_values_ages, by="Age")
p_values_all <- left_join(p_values_all, p_values_educations, by="Education")

p_values_all <- left_join(p_values_all, p_values_genders, by="Gender")

head(p_values_all)

```

```

##      Age Gender      Education assignment avg_rating
## 1: 40-50 Female Associates degree   Control      4.00
## 2: 30-40 Female Some College No degree   Control      2.75
## 3: 30-40 Female Some College No degree Treatment      2.00
## 4: 30-40 Female Bachelors degree   Control      2.25
## 5: 40-50 Female Associates degree   Control      4.00
## 6: 30-40 Female Some College No degree   Control      3.00
##      mean_rating_control.x mean_rating_treatment.x p_value.x
## 1:                3.73                3.59      0.638
## 2:                3.46                3.44      0.933

```

```
## 3:          3.46          3.44      0.933
## 4:          3.46          3.44      0.933
## 5:          3.73          3.59      0.638
## 6:          3.46          3.44      0.933
##   mean_rating_control.y mean_rating_treatment.y p_value.y mean_rating_control
## 1:          3.52          3.36      0.720          3.38
## 2:          3.23          3.43      0.285          3.38
## 3:          3.23          3.43      0.285          3.38
## 4:          3.50          3.82      0.259          3.38
## 5:          3.52          3.36      0.720          3.38
## 6:          3.23          3.43      0.285          3.38
##   mean_rating_treatment p_value
## 1:          3.47      0.535
## 2:          3.47      0.535
## 3:          3.47      0.535
## 4:          3.47      0.535
## 5:          3.47      0.535
## 6:          3.47      0.535
```

```
options(qwraps2_frmtp_digits= 2)
```

```
covariate_summary2<-
```

```
  list("Gender" =
    list("Male"      = ~ max(as.numeric(ifelse(Gender == 'Male',mean_rating_control,0))),
          "Female"    = ~ max(as.numeric(ifelse(Gender == 'Female',mean_rating_control,0)))),
    "Age" =
    list("20-30" = ~ max(as.numeric(ifelse(Age == '20-30',mean_rating_control.x,0))),
          "30-40" = ~ max(as.numeric(ifelse(Age == '30-40',mean_rating_control.x,0))),
          "40-50" = ~ max(as.numeric(ifelse(Age == '40-50',mean_rating_control.x,0))),
          "50+" = ~ max(as.numeric(ifelse(Age == '50+',mean_rating_control.x,0))),
    "Education" =
    list("Less than High school" = ~ max(as.numeric(ifelse(Education == 'Less than High school',mean_
          "High school diploma" = ~ max(as.numeric(ifelse(Education == 'High school diploma',mean_ra
          "Some College No degree" = ~ max(as.numeric(ifelse(Education == 'Some College No degree',m
          "Associates degree" = ~ max(as.numeric(ifelse(Education == 'Associates degree',mean_rating
          "Bachelors degree" = ~ max(as.numeric(ifelse(Education == 'Bachelors degree',mean_rating_co
          "Masters degree" = ~ max(as.numeric(ifelse(Education == 'Masters degree',mean_rating_contr
    )
```

```
table2<- summary_table(p_values_all , covariate_summary2)
```

```
covariate_summary3<-
```

```
  list("Gender" =
    list("Male"      = ~ max(as.numeric(ifelse(Gender == 'Male',mean_rating_treatment,0))),
          "Female"    = ~ max(as.numeric(ifelse(Gender == 'Female',mean_rating_treatment,0)))),
    "Age" =
    list("20-30" = ~ max(as.numeric(ifelse(Age == '20-30',mean_rating_treatment.x,0))),
          "30-40" = ~ max(as.numeric(ifelse(Age == '30-40',mean_rating_treatment.x,0))),
          "40-50" = ~ max(as.numeric(ifelse(Age == '40-50',mean_rating_treatment.x,0))),
          "50+" = ~ max(as.numeric(ifelse(Age == '50+',mean_rating_treatment.x,0))),
    "Education" =
    list("Less than High school" = ~ max(as.numeric(ifelse(Education == 'Less than High school',mean_
          "High school diploma" = ~ max(as.numeric(ifelse(Education == 'High school diploma',mean_ra
```



```
# Table 1: Baseline models Outcome ~ Treatment
model_1_1 <- lm(avg_rating ~ assignment, data=d)
model_1_2 <- lm(instr_rating ~ assignment, data=d)
model_1_3 <- lm(instr_subject ~ assignment, data=d)
model_1_4 <- lm(instr_material ~ assignment, data=d)
model_1_5 <- lm(instr_enthusiasm ~ assignment, data=d)
```

```
# Table 1
stargazer(model_1_1,model_1_2, model_1_3, model_1_4, model_1_5, type='text')
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               avg_rating instr_rating instr_subject instr_material instr_enthusiasm
##                               (1)         (2)         (3)         (4)         (5)
## -----
## assignmentTreatment          0.122         -0.097         0.068         0.176         0.364**
##                               (0.114)        (0.135)        (0.122)        (0.137)        (0.152)
##
## Constant                     3.450***      3.600***      3.860***      3.630***      2.780***
##                               (0.082)      (0.096)      (0.087)      (0.098)      (0.109)
## -----
## Observations                 221           221           221           221           221
## R2                           0.005         0.002         0.001         0.007         0.025
## Adjusted R2                  0.001        -0.002        -0.003         0.003         0.021
## Residual Std. Error (df = 219) 0.849         1.000         0.907         1.020         1.130
## F Statistic (df = 1; 219)      1.140         0.521         0.311         1.640         5.710**
## =====
## Note:                               *p<0.1; **p<0.05; ***p<0.01
```

```
# Table 2: Outcome with Pre-treatment blocks (Age and Gender)
model_2_1 <- lm(avg_rating ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_2_2 <- lm(instr_rating ~ as.factor(Age) + as.factor(Gender)+ assignment, data=d)
model_2_3 <- lm(instr_subject ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_2_4 <- lm(instr_material ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_2_5 <- lm(instr_enthusiasm ~ as.factor(Age) + as.factor(Gender)+assignment , data=d)
```

```
# Table 2
stargazer(model_2_1,model_2_2, model_2_3, model_2_4, model_2_5, type='text')
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               avg_rating instr_rating instr_subject instr_material instr_enthusiasm
##                               (1)         (2)         (3)         (4)         (5)
## -----
## as.factor(Age)30-40          -0.118        -0.109        -0.212        -0.098         0.043
##                               (0.159)        (0.188)        (0.170)        (0.192)        (0.208)
##
```

```
## as.factor(Age)40-50          -0.094      0.097      -0.049      0.002      -0.306
##                             (0.177)      (0.210)      (0.190)      (0.213)      (0.232)
##
## as.factor(Age)50+           -0.131      0.003      0.064      -0.035      -0.144
##                             (0.178)      (0.211)      (0.191)      (0.215)      (0.233)
##
## as.factor(Gender)Male        0.229*      0.237*      0.108      0.220      0.479***
##                             (0.120)      (0.141)      (0.128)      (0.144)      (0.156)
##
## assignmentTreatment          0.133      -0.085      0.069      0.185      0.369**
##                             (0.115)      (0.135)      (0.123)      (0.138)      (0.150)
##
## Constant                     3.450***    3.530***    3.900***    3.590***    2.680***
##                             (0.149)      (0.176)      (0.159)      (0.179)      (0.195)
##
## -----
## Observations                 221        221        221        221        221
## R2                           0.026        0.022        0.021        0.021        0.081
## Adjusted R2                  0.003       -0.001       -0.002       -0.002        0.059
## Residual Std. Error (df = 215) 0.848        1.000        0.907        1.020        1.110
## F Statistic (df = 5; 215)      1.150        0.962        0.913        0.902        3.770***
## =====
## Note:                                *p<0.1; **p<0.05; ***p<0.01
```

```
# Table 3: Outcome with Pre-treatment blocks (Age and Gender) and interaction term (gender*treatment)
model_3_1 <- lm(avg_rating ~ as.factor(Age) + as.factor(Gender)+assignment +as.factor(Gender)*assignment)
model_3_2 <- lm(instr_rating ~ as.factor(Age) + as.factor(Gender)+ assignment +as.factor(Gender)*assignment)
model_3_3 <- lm(instr_subject ~ as.factor(Age) + as.factor(Gender)+assignment+as.factor(Gender)*assignment)
model_3_4 <- lm(instr_material ~ as.factor(Age) + as.factor(Gender)+assignment+as.factor(Gender)*assignment)
model_3_5 <- lm(instr_enthusiasm ~ as.factor(Age) + as.factor(Gender)+assignment+as.factor(Gender)*assignment)
```

```
# Table 3
stargazer(model_3_1,model_3_2, model_3_3, model_3_4, model_3_5, type='text')
```

```
##
## =====
##                                     Dependent variable:
##                                     -----
##                                     avg_rating instr_rating instr_subject instr_material instr_enthusiasm
##                                     (1)          (2)          (3)          (4)          (5)
## -----
## as.factor(Age)30-40          -0.116      -0.108      -0.210      -0.092      -0.144
##                             (0.160)      (0.189)      (0.171)      (0.191)      (0.203)
##
## as.factor(Age)40-50          -0.087      0.100      -0.040      0.029      -0.001
##                             (0.179)      (0.211)      (0.191)      (0.214)      (0.233)
##
## as.factor(Age)50+           -0.129      0.004      0.067      -0.026      -0.001
##                             (0.179)      (0.211)      (0.191)      (0.214)      (0.233)
##
## as.factor(Gender)Male        0.180      0.219      0.042      0.023      0.001
##                             (0.170)      (0.201)      (0.181)      (0.203)      (0.223)
```

```
## assignmentTreatment          0.098      -0.097      0.022      0.045      0
##                             (0.143)    (0.169)    (0.153)    (0.172)    (
##
## as.factor(Gender)Male:assignmentTreatment  0.097      0.035      0.132      0.393      .
##                             (0.240)    (0.283)    (0.256)    (0.287)    (
##
## Constant          3.470***    3.540***    3.920***    3.650***    2
##                             (0.154)    (0.182)    (0.165)    (0.185)    (
##
## -----
## Observations          221      221      221      221
## R2          0.027      0.022      0.022      0.029
## Adjusted R2      -0.001      -0.005      -0.005      0.002
## Residual Std. Error (df = 214)      0.850      1.000      0.909      1.020
## F Statistic (df = 6; 214)      0.981      0.800      0.802      1.070      3
## =====
## Note:                                     *p<0.1; **p<0.05
```

Table 4:

base model

```
model_4_1 <- lm(quiz_avg~ assignment, data=d)
```

```
model_4_2 <- lm(quiz_avg ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
```

```
model_4_3 <- lm(quiz_avg ~ as.factor(Age) + as.factor(Gender)+assignment +as.factor(Gender)*assignment,
```

#models with blocking across quiz averages

```
stargazer(model_4_1, model_4_2,model_4_3, type = 'text')
```

```
##
```

```
## =====
```

```
##                                     Dependent variable:
```

```
## -----
```

	(1)	quiz_avg (2)	(3)
as.factor(Age)30-40		-0.032 (0.046)	-0.032 (0.046)
as.factor(Age)40-50		0.032 (0.051)	0.031 (0.052)
as.factor(Age)50+		0.095* (0.052)	0.095* (0.052)
as.factor(Gender)Male		-0.052 (0.035)	-0.045 (0.049)
assignmentTreatment	-0.020 (0.034)	-0.024 (0.033)	-0.019 (0.042)
as.factor(Gender)Male:assignmentTreatment			-0.013 (0.070)

## Constant	0.449***	0.454***	0.452***
##	(0.024)	(0.043)	(0.045)
##			
## -----			
## Observations	221	221	221
## R2	0.002	0.047	0.048
## Adjusted R2	-0.003	0.025	0.021
## Residual Std. Error	0.250 (df = 219)	0.246 (df = 215)	0.247 (df = 214)
## F Statistic	0.350 (df = 1; 219)	2.140* (df = 5; 215)	1.780 (df = 6; 214)
## =====			
## Note:		*p<0.1; **p<0.05; ***p<0.01	