

# Final 241 Project Consolidation

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```
# Import data and remove invalid age
```

```
d <- fread("survey_cleaned_with_states.csv")
head(d)
```

```
##           ResponseId Gender   Age           Education LocationLatitude
## 1: R_2WC5v1B2DCvHHrH   Male 30-40       Bachelors degree       52.36690
## 2: R_x2RQnFD8LZIZuRH Female 40-50       Associates degree       36.99040
## 3: R_09f8Af9asil81DH Female 30-40 Some College No degree       27.99290
## 4: R_3oTqFe6v814jV0p Female 30-40 Some College No degree       36.83070
## 5: R_0ePUcdKMDE2jNe1 Female 30-40       Bachelors degree       39.30370
## 6: R_2owAeGSUxdN9fQR Female 40-50       Associates degree       44.14149
##           LocationLongitude instr_enthusiasm instr_professional instr_subject
## 1:           4.951706                3                4                4
## 2:          -85.923897                4                4                4
## 3:          -82.403000                1                4                3
## 4:          -76.314598                2                2                2
## 5:          -94.932198                3                1                1
## 6:         -103.205200                4                4                5
##           instr_material instr_rating                subject_1
## 1:           4                4                All of the above
## 2:           4                4                Web Applications
## 3:           2                4                Web Applications
## 4:           2                2                Web Applications
## 5:           3                2 Artificial intelligence projects
## 6:           4                4                Web Applications
##                               subject_2
## 1:                               Both A & B
## 2:                               Both A & B
## 3: C. A low level implementation programming language
## 4:           B. A flexible programming language
## 5:           B. A flexible programming language
## 6:           B. A flexible programming language
##
## 1:                               Python is an object
## 2: Python is a compiled programming language with a faster and more efficient execution time than in
## 3:                               Python is an object
## 4:                               Python was crea
## 5:                               Python was crea
## 6:                               Python is an objec
##           subject_4           video_watched assigned assignment ques_1
## 1:           Easy to Learn Video1-1(Treatment)                1 Treatment                1
```

```
## 2: Great Advanced Language Video3-2(Control) 0 Control 0
## 3: Easy to Learn Video2-2(Control) 0 Control 0
## 4: Wonderful Community Video2-1(Treatment) 1 Treatment 0
## 5: Wonderful Community Video1-2(Control) 0 Control 0
## 6: Wonderful Community Video2-2(Control) 0 Control 0
## ques_2 ques_3 ques_4 quiz_avg avg_rating state state_code dem_percent
## 1: 1 0 0 0.50 3.75 NA
## 2: 1 1 1 0.75 4.00 kentucky KY 0.362
## 3: 0 0 0 0.00 2.75 florida FL 0.479
## 4: 0 0 0 0.00 2.00 virginia VA 0.541
## 5: 0 0 0 0.00 2.25 kansas KS 0.416
## 6: 0 0 0 0.00 4.00 south dakota SD 0.356
## rep_percent
## 1: NA
## 2: 0.621
## 3: 0.512
## 4: 0.440
## 5: 0.562
## 6: 0.618
```

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

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)]
length(d$college_degree)
```

```
## [1] 222
```

## 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 15
## 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.
```

<b>**Characteristic**</b>	<b>**Overall**</b> , N = 222	<b>**Female**</b> , N = 142	<b>**Male**</b> , N = 80
assignment			
Control	108 (49%)	68 (48%)	40 (50%)
Treatment	114 (51%)	74 (52%)	40 (50%)
Age			
20-30	44 (20%)	25 (18%)	19 (24%)
30-40	83 (37%)	56 (39%)	27 (34%)
40-50	48 (22%)	32 (23%)	16 (20%)
50+	47 (21%)	29 (20%)	18 (22%)

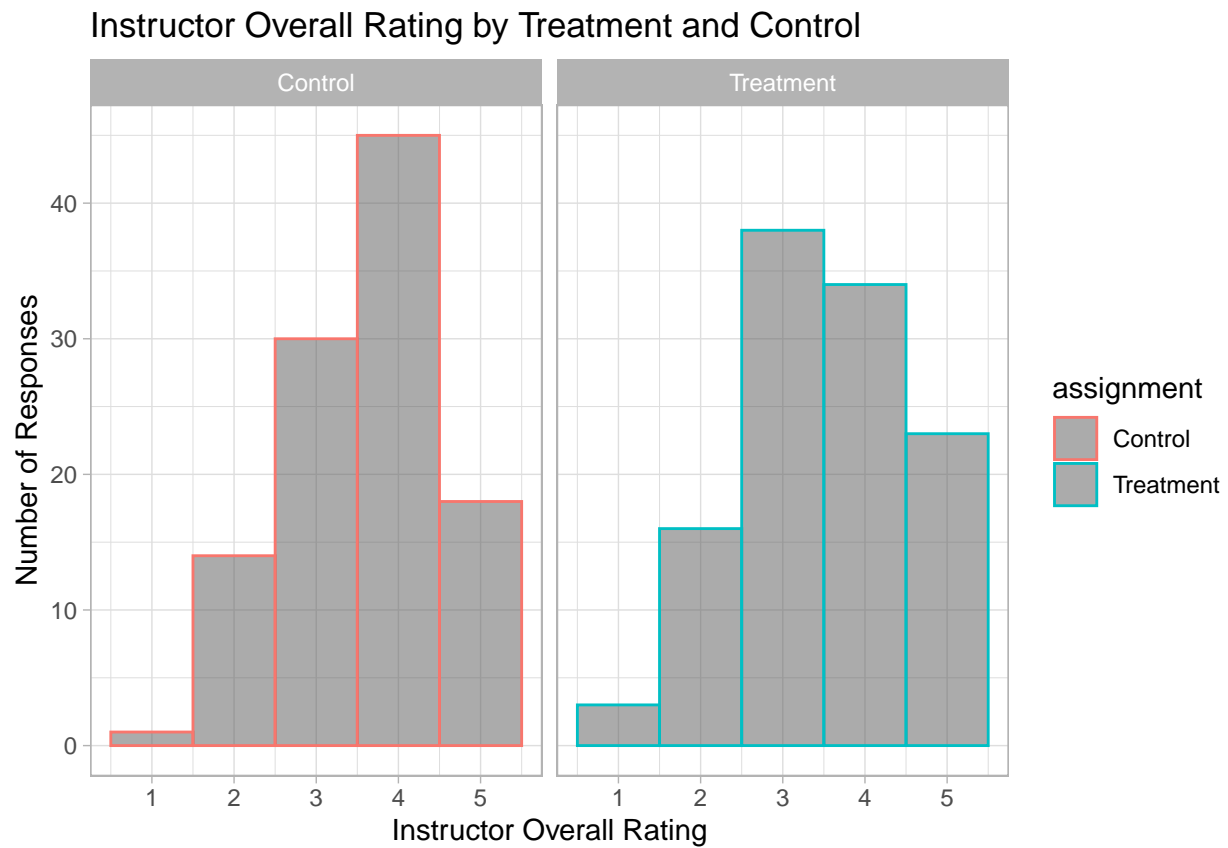
```
d %>%
  select(Gender, Education, assignment) %>%
  tbl_summary(by = assignment) %>%
  add_overall()
```

```
## 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.
```

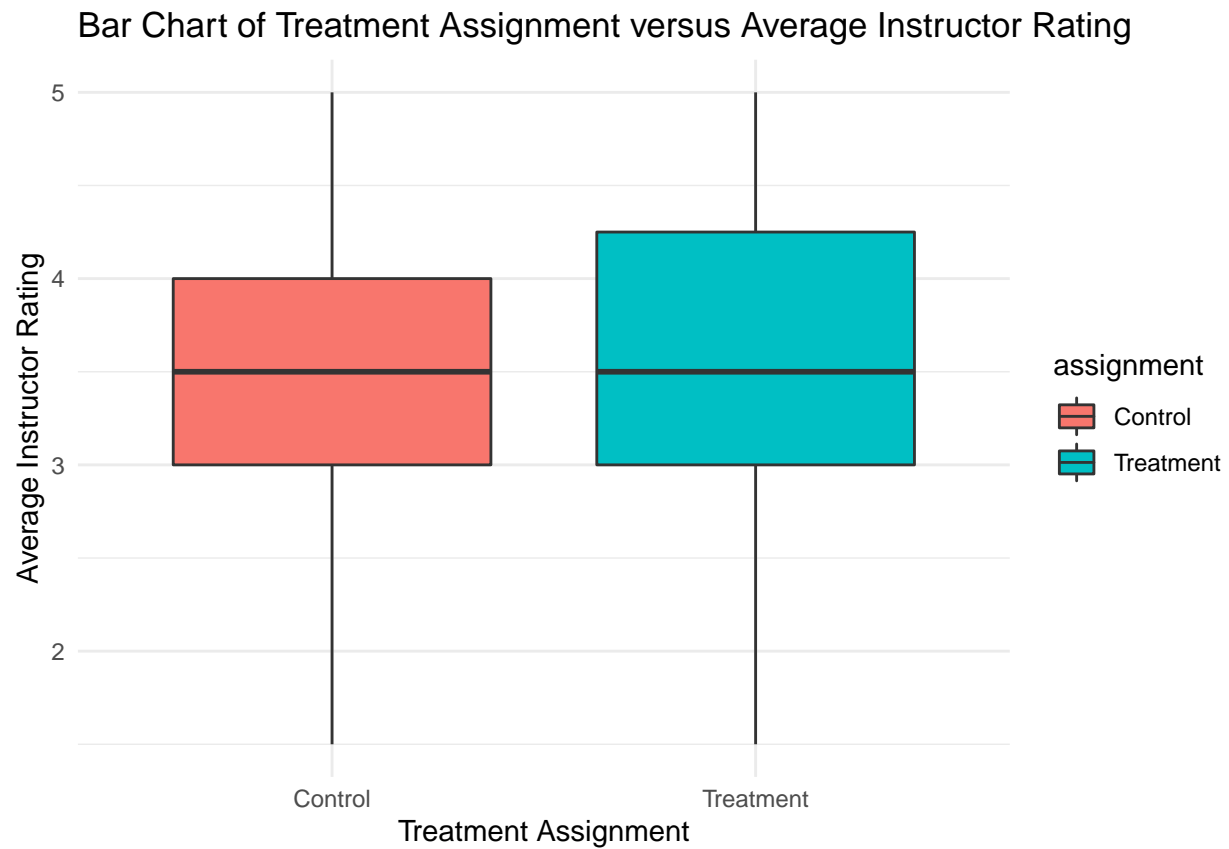
<b>**Characteristic**</b>	<b>**Overall**</b> , N = 222	<b>**Control**</b> , N = 108	<b>**Treatment**</b> , N = 114
Gender			
Female	142 (64%)	68 (63%)	74 (65%)
Male	80 (36%)	40 (37%)	40 (35%)
Education			
Associates degree	27 (12%)	16 (15%)	11 (9.6%)
Bachelors degree	44 (20%)	21 (19%)	23 (20%)
High school 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)
```

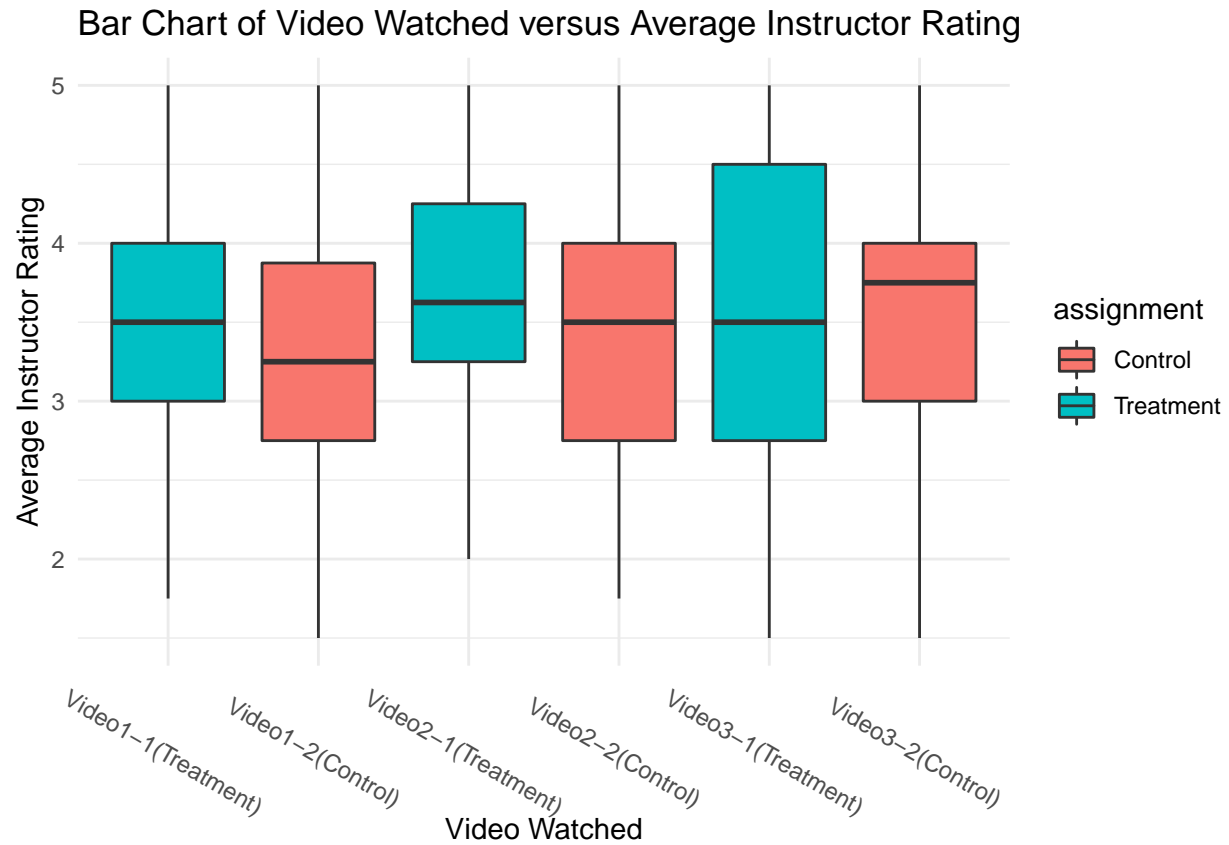
response\_hist



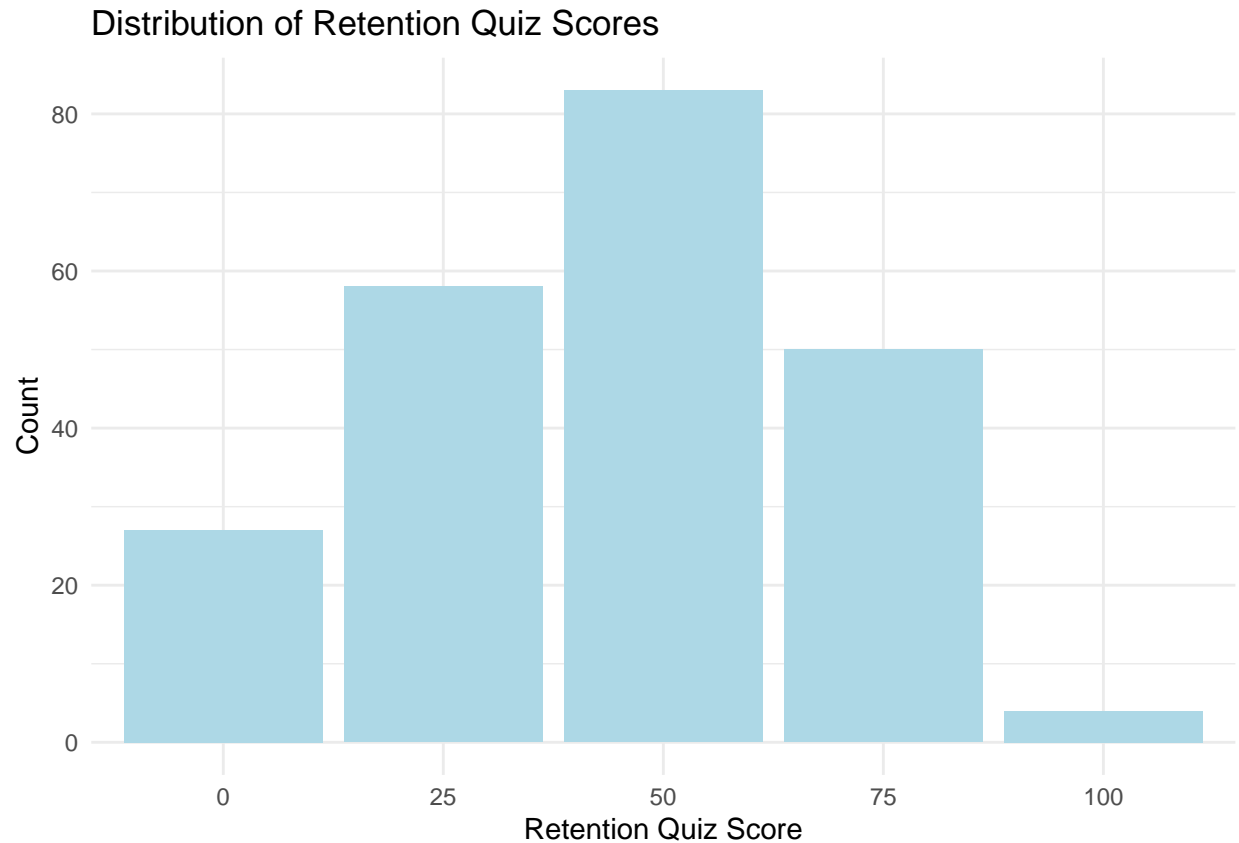
```
ggplot(d, aes(x=assignment, y=avg_rating, fill=assignment))+  
  geom_boxplot()+  
  labs(title="Bar Chart of Treatment Assignment versus Average Instructor Rating",  
        x="Treatment Assignment", y="Average Instructor Rating")
```



```
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     217 55.036
## 2     221 55.459 -4   -0.42338 0.4173 0.7961

```

```

# Separate data objects for treatment and control groups
d_treat <-
  d %>%
  filter(assignment == 'Treatment')

d_control <-

```

```

d %>%
  filter(assignment == 'Control')

# 2 sided t-test for each one of the instructor questions
instructor_enthusiams_ttest <- t.test(d_treat$instr_enthusiasm, d_control$instr_enthusiasm, conf.level = 0.95)
instructor_enthusiams_ttest

##
## Welch Two Sample t-test
##
## data: d_treat$instr_enthusiasm and d_control$instr_enthusiasm
## t = 2.385, df = 214.8, p-value = 0.01795
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  0.0629284 0.6622178
## sample estimates:
## mean of x mean of y
##  3.140351  2.777778

instructor_professional_ttest <- t.test(d_treat$instr_professional, d_control$instr_professional, conf.level = 0.95)
instructor_professional_ttest

##
## Welch Two Sample t-test
##
## data: d_treat$instr_professional and d_control$instr_professional
## t = 0.38053, df = 213.09, p-value = 0.7039
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1955552 0.2891224
## sample estimates:
## mean of x mean of y
##  3.824561  3.777778

instructor_subject_ttest <- t.test(d_treat$instr_subject, d_control$instr_subject, conf.level = 0.95)
instructor_subject_ttest

##
## Welch Two Sample t-test
##
## data: d_treat$instr_subject and d_control$instr_subject
## t = 0.56564, df = 219.88, p-value = 0.5722
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1706972 0.3081241
## sample estimates:
## mean of x mean of y
##  3.929825  3.861111

```



```
instructor_material_ttest <- t.test(d_treat$instr_material, d_control$instr_material, conf.level = 0.95)
instructor_material_ttest
```

```
##
## Welch Two Sample t-test
##
## data: d_treat$instr_material and d_control$instr_material
## t = 1.3029, df = 219.98, p-value = 0.194
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.0909308 0.4457066
## sample estimates:
## mean of x mean of y
## 3.807018 3.629630
```

```
instructor_overall_ttest <- t.test(d_treat$instr_rating, d_control$instr_rating, conf.level = 0.95)
instructor_overall_ttest
```

```
##
## Welch Two Sample t-test
##
## data: d_treat$instr_rating and d_control$instr_rating
## t = -0.69453, df = 219.48, p-value = 0.4881
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.3572076 0.1710477
## sample estimates:
## mean of x mean of y
## 3.508772 3.601852
```

```
model_0 <- lm(avg_rating ~ assignment, data=d)
stargazer(model_0, type='text')
```

```
##
## =====
##                      Dependent variable:
##                      -----
##                      avg_rating
## -----
## assignmentTreatment      0.123
##                          (0.114)
##
## Constant                 3.447***
##                          (0.082)
##
## -----
## Observations              222
## R2                        0.005
## Adjusted R2               0.001
## Residual Std. Error      0.848 (df = 220)
## F Statistic               1.176 (df = 1; 220)
## =====
## Note:                    *p<0.1; **p<0.05; ***p<0.01
```

```

model_1 <- lm(avg_rating ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_2 <- lm(instr_rating ~ as.factor(Age) + as.factor(Gender)+ assignment, data=d)
model_3 <- lm(instr_subject ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_4 <- lm(instr_material ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_5 <- lm(instr_enthusiasm ~ as.factor(Age) + as.factor(Gender)+assignment , data=d)
model_6 <- lm(instr_enthusiasm ~ as.factor(Age) + as.factor(Gender)+assignment +as.factor(Gender)*assignment,
model_7 <- lm(avg_rating ~ as.factor(Age) + as.factor(Gender)+assignment +as.factor(Gender)*assignment,
model_8 <- lm(instr_rating ~ as.factor(Age) + as.factor(Gender)+assignment +as.factor(Gender)*assignment

```

```
#models with blocking across instructor ratings
```

```
stargazer(model_1, model_2, model_3,model_4, model_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              -0.117      -0.103      -0.210      -0.096      0.035
##                                (0.158)      (0.187)      (0.169)      (0.191)      (0.207)
##
## as.factor(Age)40-50              -0.094       0.098      -0.048       0.003     -0.307
##                                (0.177)      (0.209)      (0.189)      (0.213)      (0.232)
##
## as.factor(Age)50+                -0.131       0.003       0.064      -0.035     -0.144
##                                (0.178)      (0.210)      (0.190)      (0.214)      (0.233)
##
## as.factor(Gender)Male             0.230*       0.243*       0.110       0.221     0.472***
##                                (0.119)      (0.140)      (0.127)      (0.143)      (0.155)
##
## assignmentTreatment               0.133       -0.081       0.070       0.186     0.364**
##                                (0.114)      (0.135)      (0.122)      (0.137)      (0.149)
##
## Constant                         3.451***     3.525***     3.896***     3.589***     2.691***
##                                (0.148)      (0.175)      (0.159)      (0.178)      (0.194)
##
## -----
## Observations                     222          222          222          222          222
## R2                               0.026          0.022          0.021          0.021          0.079
## Adjusted R2                     0.004          -0.001         -0.002         -0.002          0.058
## Residual Std. Error (df = 216)  0.846          1.000          0.905          1.018          1.107
## F Statistic (df = 5; 216)       1.170          0.970          0.915          0.919          3.726***
## =====
## Note:                               *p<0.1; **p<0.05; ***p<0.01

```

```
#interaction term models
```

```
stargazer( model_6, model_7, model_8, type = 'text')
```

```

##
## =====
##                                     Dependent variable:

```

```
## -----
## instr_enthusiasm avg_rating instr_rating
## (1) (2) (3)
## -----
## as.factor(Age)30-40 0.034 -0.116 -0.102
## (0.208) (0.159) (0.188)
##
## as.factor(Age)40-50 -0.315 -0.087 0.101
## (0.233) (0.178) (0.210)
##
## as.factor(Age)50+ -0.147 -0.129 0.004
## (0.233) (0.178) (0.211)
##
## as.factor(Gender)Male 0.536** 0.180 0.220
## (0.222) (0.169) (0.200)
##
## assignmentTreatment 0.409** 0.098 -0.097
## (0.187) (0.143) (0.169)
##
## as.factor(Gender)Male:assignmentTreatment -0.127 0.098 0.045
## (0.312) (0.238) (0.282)
##
## Constant 2.670*** 3.467*** 3.532***
## (0.201) (0.154) (0.181)
## -----
## Observations 222 222 222
## R2 0.080 0.027 0.022
## Adjusted R2 0.054 -0.00003 -0.005
## Residual Std. Error (df = 215) 1.110 0.848 1.002
## F Statistic (df = 6; 215) 3.120*** 0.999 0.809
## =====
## Note: *p<0.1; **p<0.05; ***p<0.01
```

```
# base model
model_0_2 <- lm(quiz_avg~ assignment, data=d)
model_1_2 <- lm(quiz_avg ~ as.factor(Age) + as.factor(Gender)+assignment, data=d)
model_2_2 <- lm(quiz_avg ~ as.factor(Age) + as.factor(Gender)+assignment +as.factor(Gender)*assignment,
```

```
#models with blocking across quiz averages
stargazer(model_1_2, model_2_2, type = 'text')
```

```
##
## =====
## Dependent variable:
## -----
## quiz_avg
## (1) (2)
## -----
## as.factor(Age)30-40 -0.030 -0.030
## (0.046) (0.046)
##
## as.factor(Age)40-50 0.033 0.032
```

##	(0.051)	(0.052)
##		
## as.factor(Age)50+	0.095*	0.095*
##	(0.052)	(0.052)
##		
## as.factor(Gender)Male	-0.050	-0.045
##	(0.034)	(0.049)
##		
## assignmentTreatment	-0.023	-0.019
##	(0.033)	(0.041)
##		
## as.factor(Gender)Male:assignmentTreatment		-0.010
##		(0.069)
##		
## Constant	0.453***	0.451***
##	(0.043)	(0.045)
##		
## -----		
## Observations	222	222
## R2	0.046	0.046
## Adjusted R2	0.024	0.019
## Residual Std. Error	0.246 (df = 216)	0.246 (df = 215)
## F Statistic	2.080* (df = 5; 216)	1.728 (df = 6; 215)
## =====		
## Note:	*p<0.1; **p<0.05; ***p<0.01	