# 590 HW1

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## 4/13/2021

```
# Load packages
library(pacman)
p_load(readr, data.table, dplyr, janitor, haven, here,
       tidyverse, skimr, lfe, stargazer, quantreg, hrbrthemes, tinytex,kableExtra)
star = stargazer(attitude, header = F)
##
## \begin{table}[!htbp] \centering
     \caption{}
     \label{}
##
## \begin{tabular}{@{\extracolsep{5pt}}lccccccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
## Statistic & \multicolumn{1}{c}{N} & \multicolumn{1}{c}{Mean} & \multicolumn{1}{c}{St. Dev.} & \multicolumn{1} & \multicolumn{1}{c}{St. Dev.} & \multicolumn{1}{c}
## \hline \\[-1.8ex]
## rating & 30 & 64.633 & 12.173 & 40 & 58.8 & 71.8 & 85 \\
## complaints & 30 & 66.600 & 13.315 & 37 & 58.5 & 77 & 90 \\
## privileges & 30 & 53.133 & 12.235 & 30 & 45 & 62.5 & 83 \\
## learning & 30 & 56.367 & 11.737 & 34 & 47 & 66.8 & 75 \\
## raises & 30 & 64.633 & 10.397 & 43 & 58.2 & 71 & 88 \\
## critical & 30 & 74.767 & 9.895 & 49 & 69.2 & 80 & 92 \\
## advance & 30 & 42.933 & 10.289 & 25 & 35 & 47.8 & 72 \\
## \hline \\[-1.8ex]
## \end{tabular}
## \end{table}
star = sub('^.+\\caption.+$','', star)
# Load data
af = read_dta(here("afghanistan_anonymized_data.dta"))
# Create nonoutlier column
af <- af %>% mutate(nonoutlier= ifelse((f07_num_ppl_hh_cnt > 20 & f07_observed == 1) |
                                      (f07_jeribs_cnt > 10 \& f07_observed == 1)
                                      (f07_num_sheep_cnt > 50 \& f07_observed == 1)
                                        (s08_num_ppl_hh_cnt > 20 \& s08_observed == 1)
                                        (s08_jeribs_cnt > 10 & s08_observed == 1) |
                                        (s08_num_sheep_cnt > 50 & s08_observed == 1)
                                        ,0,1))
```

```
# Tabulating nonoutliers
non_outlier_tab <- af %>% group_by(nonoutlier) %>%
    summarise(count = n(), percentage = n()/nrow(.))
non_outlier_tab
## # A tibble: 2 x 3
   nonoutlier count percentage
          <dbl> <int>
                           <dbl>
## 1
              0
                   76
                          0.0421
## 2
              1 1728
                          0.958
There are 76 non-outliers in the data given the specifications prior.
# Run girls regression table 4 column 1 including outliers
reg3 = felm(f07_formal_school ~ treatment + chagcharan | 0 | 0 | clustercode,
            data = af %>% filter(f07_girl_cnt == 1 &
                                 f07_{observed} == 1))
# Run boys regression table 4 column lincluding outliers
reg4 = felm(f07_formal_school ~ treatment + chagcharan | 0 | 0 | clustercode,
            data = af %>% filter(f07_girl_cnt == 0 &
                                 f07_{observed} == 1))
# Run girls regression table 4 column 1 with controls
reg5 = felm(f07_formal_school ~ treatment + chagcharan + f07_heads_child_cnt +
              f07_age_cnt + f07_duration_village_cnt +
              f07_farsi_cnt + f07_tajik_cnt + f07_farmer_cnt +
              f07_age_head_cnt + f07_yrs_ed_head_cnt + f07_num_ppl_hh_cnt +
              f07_jeribs_cnt + f07_num_sheep_cnt +
              f07_nearest_scl| 0 | 0 | clustercode,
            data = af %>% filter(f07_girl_cnt == 1 &
                                 nonoutlier == 1 &
                                 f07_{observed} == 1))
# Run girls regression table 4 column 1 with controls
reg6 = felm(f07_formal_school ~ treatment + chagcharan + f07_heads_child_cnt +
              f07 age cnt + f07 duration village cnt +
              f07_farsi_cnt + f07_tajik_cnt + f07_farmer_cnt +
              f07_age_head_cnt + f07_yrs_ed_head_cnt + f07_num_ppl_hh_cnt +
              f07_jeribs_cnt + f07_num_sheep_cnt +
              f07_nearest_scl| 0 | 0 | clustercode,
```

### Question 3:

nonoutlier == 1 &
f07\_observed == 1))

data = af %>% filter(f07\_girl\_cnt == 0 &

# Question 4:

```
# girls regression column 4
reg9 = felm(f07_both_norma_total ~ treatment + chagcharan + f07_heads_child_cnt +
              f07_age_cnt + f07_duration_village_cnt +
              f07_farsi_cnt + f07_tajik_cnt + f07_farmer_cnt +
              f07_age_head_cnt + f07_yrs_ed_head_cnt + f07_num_ppl_hh_cnt +
              f07_jeribs_cnt + f07_num_sheep_cnt +
              f07_nearest_scl| 0 | 0 | clustercode,
            data = af %>% filter(f07_girl_cnt == 1 &
                                 nonoutlier == 1 &
                                 f07_test_observed == 1))
# boys regression column 4
reg10 = felm(f07_both_norma_total ~ treatment + chagcharan + f07_heads_child_cnt +
              f07_age_cnt + f07_duration_village_cnt +
              f07_farsi_cnt + f07_tajik_cnt + f07_farmer_cnt +
              f07_age_head_cnt + f07_yrs_ed_head_cnt + f07_num_ppl_hh_cnt +
              f07_jeribs_cnt + f07_num_sheep_cnt +
              f07_nearest_scl| 0 | 0 | clustercode,
            data = af %>% filter(f07_girl_cnt == 0 &
                                 nonoutlier == 1 &
                                 f07_test_observed == 1))
```

# Question 4b:

```
s08_test_observed == 1))
# girls regression for column 7
reg13 = felm(s08_both_norma_total ~ treatment + chagcharan +
               s08_heads_child_cnt + s08_age_cnt +
               s08_duration_village_cnt + s08_farsi_cnt + s08_tajik_cnt +
               s08_farmer_cnt + s08_age_head_cnt + s08_yrs_ed_head_cnt +
               s08_num_ppl_hh_cnt + s08_jeribs_cnt + s08_num_sheep_cnt +
               s08_nearest_scl | 0 | 0 | clustercode,
            data = af %>% filter(s08 girls cnt == 1 &
                                 nonoutlier == 1 &
                                 s08 test observed == 1))
# boys regression for column 7
reg14 = felm(s08_both_norma_total ~ treatment + chagcharan +
               s08_heads_child_cnt + s08_age_cnt +
               s08_duration_village_cnt + s08_farsi_cnt + s08_tajik_cnt +
               s08_farmer_cnt + s08_age_head_cnt + s08_yrs_ed_head_cnt +
               s08_num_ppl_hh_cnt + s08_jeribs_cnt + s08_num_sheep_cnt +
               s08_nearest_scl | 0 | 0 | clustercode,
            data = af %>% filter(s08_girls_cnt == 0 &
                                 nonoutlier == 1 &
                                 s08_test_observed == 1))
# Getting mean values for column 1
col_1 = af %>%
  filter(treatment == 1 & nonoutlier == 1 & f07_observed == 1) %>%
  select(
    c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
                   f07_num_sheep_cnt, f07_nearest_scl)
  ) %>%
  mutate(
  across(.cols = c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
                   f07_num_sheep_cnt, f07_nearest_scl), mean, na.rm = TRUE)
  ) %>% head(., 1) %>% unlist(.)
# Getting mean values for column 2
col_2 = af \%
 filter(treatment == 0 & nonoutlier == 1 & f07_observed == 1) %>%
  select(
    c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
                   f07_num_sheep_cnt, f07_nearest_scl)
  ) %>%
  mutate(
  across(.cols = c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
```

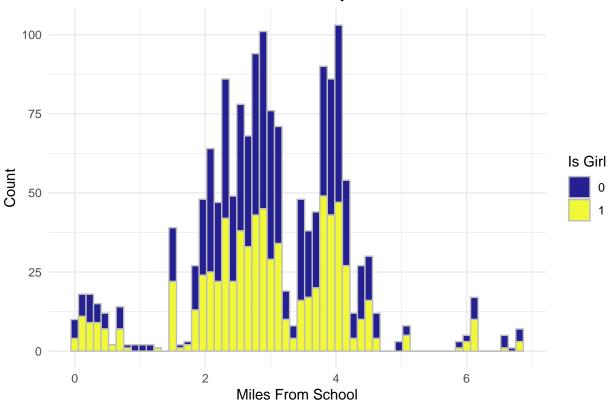
nonoutlier == 1 &

```
f07_num_sheep_cnt, f07_nearest_scl), mean, na.rm = TRUE)
  ) %>% head(., 1) %>% unlist(.)
# Create regression functions for column 3
reg_function_coef = function(x){
    reg = felm(x ~ treatment | 0 | 0 | clustercode,
               data = af %>% filter(nonoutlier == 1 & f07 observed == 1))
    coef = coef(reg)[2]
    return(coef)
}
# Regression function for SE
reg_function_se = function(x){
  reg = felm(x ~ treatment | 0 | 0 | clustercode,
             data = af %>% filter(nonoutlier == 1 & f07_observed == 1))
  se = sqrt(diag(vcov(reg)))[2]
  return(se)
# Obtain column 3 coefficients
col 3 coefs = af \%
  filter(nonoutlier == 1 & f07 observed == 1) %>%
  select(
    c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
                   f07_num_sheep_cnt, f07_nearest_scl)
  ) %>%
  mutate(
  across(.cols = c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
                   f07_num_sheep_cnt, f07_nearest_scl), reg_function_coef)
  ) %>% head(., 1) %>% unlist(.)
# Obtain column 3 SE
col_3_se = af %>%
  filter(nonoutlier == 1 & f07_observed == 1) %>%
  select(
    c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
                   f07_num_sheep_cnt, f07_nearest_scl)
  ) %>%
  mutate(
  across(.cols = c(f07_heads_child_cnt, f07_girl_cnt, f07_age_cnt, f07_duration_village_cnt,
                   f07_farsi_cnt, f07_tajik_cnt, f07_farmer_cnt,
                   f07_age_head_cnt, f07_yrs_ed_head_cnt, f07_num_ppl_hh_cnt, f07_jeribs_cnt,
```

```
f07_num_sheep_cnt, f07_nearest_scl), reg_function_se)
  ) %>% head(., 1) %>% unlist(.)
# Regression for column 7
col_7 = felm(f07_formal_school ~ f07_heads_child_cnt +
               f07_girl_cnt + f07_age_cnt + f07_duration_village_cnt +
               f07_farsi_cnt + f07_tajik_cnt + f07_farmer_cnt +
               f07_age_head_cnt + f07_yrs_ed_head_cnt + f07_num_ppl_hh_cnt +
               f07_jeribs_cnt + f07_num_sheep_cnt + f07_nearest_scl
             | 0 | 0 | clustercode,
             data = af %>%
               filter(nonoutlier == 1 & f07_observed == 1 & treatment == 0))
# Regression for column 8
col_8 = felm(f07_both_norma_total ~ f07_heads_child_cnt +
               f07_girl_cnt + f07_age_cnt + f07_duration_village_cnt +
               f07_farsi_cnt + f07_tajik_cnt + f07_farmer_cnt +
               f07_age_head_cnt + f07_yrs_ed_head_cnt + f07_num_ppl_hh_cnt +
               f07_jeribs_cnt + f07_num_sheep_cnt + f07_nearest_scl
             | 0 | 0 | clustercode,
             data = af \%
               filter(nonoutlier == 1 & f07_observed == 1 &
                        f07_test_observed == 1 & treatment == 0))
# further analysis regression
reg_fa = felm(f07_formal_school ~ f07_girl_cnt +
         f07_nearest_scl + f07_girl_cnt*f07_nearest_scl +
         I(f07 nearest scl^2) | 0 | 0 | clustercode,
         data = af %>% filter(treatment==0 &
                              f07_observed == 1 &
                              nonoutlier == 1))
ggplot(af, aes(x = f07_nearest_scl, fill = as.factor(f07_girl_cnt))) +
  geom_histogram(color = "grey", bins = 60, alpha = .9) +
  scale_fill_viridis_d(option = "C") +
  theme_minimal() +
  labs(x = "Miles From School", y = "Count", fill = "Is Girl",
       title = "Distribution of Distance from School By Gender")
```

## Warning: Removed 233 rows containing non-finite values (stat\_bin).





```
af = af %>% mutate(dist_near = if_else(f07_nearest_scl < 2, 1, 0),</pre>
               dist_close = if_else(f07_nearest_scl >= 2 & f07_nearest_scl < 3, 1, 0),</pre>
               dist_nclose = if_else(f07_nearest_scl >= 3 & f07_nearest_scl < 4, 1, 0),</pre>
               dist_far = if_else(f07_nearest_scl >= 4, 1, 0))
# Regression for near individuals
reg_near_g = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                               f07_observed == 1 &
                               nonoutlier == 1 &
                               f07_girl_cnt == 1 &
                               dist near == 1
                               ))
# Regression for close individuals
reg_close_g = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                               f07_observed == 1 &
                               nonoutlier == 1 &
                               f07_girl_cnt == 1 &
                               dist_close == 1
# Regression for not close individuals
```

```
reg_nclose_g = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 1 &
                              dist_nclose == 1
# Regression for far individuals
reg_far_g = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 1 &
                              dist_far == 1
# Regression for near individuals
reg_near_b= felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist near == 1
                              ))
# Regression for close individuals
reg_close_b = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist_close == 1
                              ))
# Regression for not close individuals
reg_nclose_b = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist_nclose == 1
# Regression for far individuals
reg_far_b = felm(f07_formal_school ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
```

```
f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist_far == 1
                              ))
# Regression for near individuals
reg_test_near_g = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 1 &
                              dist_near == 1
# Regression for close individuals
reg_test_close_g = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 1 &
                              dist_close == 1
# Regression for not close individuals
reg_test_nclose_g = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 1 &
                              dist_nclose == 1
                              ))
# Regression for far individuals
reg_test_far_g = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 1 &
                              dist_far == 1
                              ))
# Regression for near individuals
reg_test_near_b= felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
```

```
dist_near == 1
# Regression for close individuals
reg_test_close_b = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist_close == 1
                              ))
# Regression for not close individuals
reg_test_nclose_b = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist_nclose == 1
                              ))
# Regression for far individuals
reg_test_far_b = felm(f07_both_norma_total ~ treatment +
                   chagcharan | 0 | 0 | clustercode,
         data = af %>% filter(
                              f07_observed == 1 &
                              nonoutlier == 1 &
                              f07_girl_cnt == 0 &
                              dist_far == 1
# Table 4.1
stargazer(reg1, reg2, reg5, reg6, reg7, reg8, reg9, reg10,
          keep.stat = c('n', 'rsq'),
          column.labels = c(" ", "
                                             11 11
                                 ", "
                                           ", "
          title = "Table 4: Treatment Effects by Gender",
          covariate.labels = c('Treatment', "chagcharan",
                               "Household head's child", "Age",
                               "Years family in village", "Farsi",
                               "Tajik", "Farmers", "Age of household head",
                               "Years of education of household head",
                               "Jeribs of land", "Number of sheep",
                               "Distance to nearest formal school",
                               "Nearest School"),
          dep.var.labels = c("Formaly Enroled", "Fall Test Scores"),
          \#dep.var.labels = c(""),
          type = 'latex', header = F, float = TRUE,
          font.size = "small",
          column.sep.width = "-15pt",
          omit.stat=c("f", "ser"),
```

Table 1: Table 4: Treatment Effects by Gender

	Dependent variable:							
_	Formaly Enroled			Fall Test Scores			3	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Treatment				0.347*** (0.094)				
chagcharan			$0.154^{*}$ $(0.082)$		0.282** (0.123)			0.118 (0.075)
Household head's child			-0.043 $(0.051)$	0.022 $(0.051)$			-0.156 $(0.168)$	0.125 (0.098)
m Age				0.065*** (0.019)				0.367*** (0.021)
Years family in village				-0.0001 $(0.002)$				-0.003 $(0.002)$
Farsi			-0.082 $(0.051)$	0.019 (0.063)			-0.115 $(0.097)$	0.094 (0.141)
Tajik			-0.063	0.077*** (0.030)			-0.005	0.173*** (0.052)
Farmers				$-0.082^{**}$ $(0.039)$				-0.082 (0.113)
Age of household head		-	-0.00004 $(0.002)$	4-0.003 $(0.002)$			-0.001 $(0.004)$	0.005 (0.003)
Years of education of household hea	d		0.003	0.002 (0.004)			,	0.048*** (0.011)
Jeribs of land			0.007	-0.001 $(0.009)$			0.007 (0.007)	-0.001 $(0.014)$
Number of sheep			-0.009	,			0.016	0.018 (0.032)
Distance to nearest formal school			0.006	0.004 (0.003)			0.008 (0.008)	0.013**
Nearest School			-0.007	$-0.059^{**}$ $(0.024)$			0.001	-0.070 $(0.049)$
Constant			-0.168	,			-2.421**	*-3.017***
Observations R <sup>2</sup>	693	797 0.164	693 0.371	797 0.245	667 0.167	707 0.045	667 0.357	707 0.404

Table 2: Table 4b: Treatment Effects by Gender

	$Dependent\ variable:$							
	s08_both_norma_total							
	(1)	(2)	(3)	(4)				
Treatment	$0.735^{***} (0.093)$	0.380*** (0.129)	0.661*** (0.090)	$0.413^{***}$ $(0.099)$				
chagcharan	$-0.458^{***}$ $(0.066)$	0.281*** (0.089)	$-3.052^{***}$ $(0.460)$	$-3.144^{***}$ (0.302)				
Observations $\mathbb{R}^2$	689 0.165	712 0.042	687 0.378	709 0.410				

Note:

S.E Clustered by village

Table 3: Table 2: Treatment Effects by Gender

	Dependent variable:			
	Formal enrollment	Test Scores		
	(1)	(2)		
Household head's child	0.038	-0.090		
	(0.061)	(0.113)		
Girl	-0.208***	-0.682***		
	(0.080)	(0.102)		
m Age	0.046***	0.287***		
	(0.017)	(0.018)		
Years family in village	-0.001	-0.005**		
	(0.001)	(0.002)		
Farsi	-0.039	0.074		
	(0.074)	(0.119)		
Tajik	-0.006	0.042		
	(0.078)	(0.065)		
Farmers	-0.050	-0.020		
	(0.081)	(0.120)		
Age of household head	-0.004	-0.004**		
	(0.003)	(0.002)		
Years of education of household head	0.001	0.036***		
	(0.006)	(0.007)		
Number of people in household	0.004	-0.004		
	(0.006)	(0.011)		
Jeribs of land	0.022***	0.056***		
	(0.008)	(0.011)		
Number of sheep	0.010***	0.016**		
	(0.002)	(0.007)		
Distance to nearest formal school	-0.060	-0.089**		
	(0.043)	(0.037)		
Constant	0.246	-1.649***		
	(0.324)	(0.369)		
Observations	708	653		
$\frac{R^2}{R^2}$	0.155	0.401		
Note:	*p<0.1; **p<0.05; ***p<0.01			

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01 S.E Clustered by village

Table 4: Table 2

Name	Treatment Avg	Control Avg	Estimated Diff	Estimated Difference SE
Household head's child	0.9347826	0.9110169	0.0237657	0.0148157
Girl	0.4744246	0.4548023	0.0196223	0.0198052
Age	8.3209719	8.3121469	0.0088250	0.0399960
Years family in village	30.3024297	27.5939266	2.7085031	1.6051614
Farsi	0.2084399	0.2090395	-0.0005997	0.0544541
Tajik	0.2429668	0.2076271	0.0353396	0.0489753
Farmers	0.7173913	0.7274011	-0.0100098	0.0335306
Age of household head	40.1419437	39.9703390	0.1716048	1.1006431
Years of education of household head	3.3145780	3.0755650	0.2390130	0.4416666
Number of People in Household	8.3989770	7.8177966	0.5811804	0.3399672
Jeribs of land	1.3446292	1.2740113	0.0706179	0.1069723
Number of sheep	7.5524297	5.6313559	1.9210737	1.5041764
Distance to nearest formal school	2.9099495	3.1628723	-0.2529228	0.3492530

Table 5: Further Analysis: Distance Effect on Enrollment

	Dependent variable:								
	Whether or not the Individual is in a Formal School								
	Short	Med-Short	Med-Long	Long	Short	Med-Short	Med-Long	g Long	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	
Treatment	0.448***	0.528***	0.696***	0.481***	$0.319^{*}$	0.447***	0.235**	0.504***	
	(0.122)	(0.140)	(0.073)	(0.066)	(0.190)	(0.160)	(0.093)	(0.092)	
chagcharan	0.404***	0.192	-0.156	0.368***	0.377**	0.028	0.058	0.153	
	(0.069)	(0.117)	(0.126)	(0.064)	(0.150)	(0.152)	(0.111)	(0.094)	
Constant	0.128	0.022	0.072	0.049	0.357*	0.332***	0.414***	0.094	
	(0.124)	(0.026)	(0.067)	(0.046)	(0.203)	(0.063)	(0.060)	(0.084)	
Observations	s 94	272	204	123	88	328	232	149	
$\mathbb{R}^2$	0.173	0.400	0.403	0.420	0.225	0.213	0.068	0.310	

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

S.E Clustered by village 1-4 (Female), 5-8 (Male).

Both groups have increasing distance as the number increases.

Table 6: Further Analysis: Distance Effect on Test Scores

	Dependent variable:							
	Total Normalized Test Score, Fall 2007							
	Short	Med-Short	Med-Long	Long	Short	Med-Short	Med-Long	Long
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Treatment	1.126***	0.789***	0.853***	0.563***	$0.103^{*}$	0.512***	0.202*	0.895***
	(0.262)	(0.233)	(0.198)	(0.065)	(0.061)	(0.159)	(0.116)	(0.152)
chagcharan	0.773***	$0.360^{*}$	-0.277	0.445***	-0.027	0.146	-0.111	0.371***
	(0.248)	(0.184)	(0.197)	(0.066)	(0.053)	(0.114)	(0.194)	(0.121)
Constant	-1.169***	-0.496***	-0.399**	-0.559***	0.396***	0.281***	0.497***	-0.351***
	(0.295)	(0.134)	(0.177)	(0.070)	(0.048)	(0.040)	(0.121)	(0.080)
Observations	91	259	196	121	81	290	198	138
$\frac{R^2}{R^2}$	0.217	0.212	0.174	0.193	0.002	0.069	0.006	0.221

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

S.E Clustered by village 1-4 (Female), 5-8 (Male).

Both groups have increasing distance as the number increases.

#### **#Written Questions:**

##Outlier question 1 An outlier observation is someone who had more than 50 sheep or goats owned by the household in 2007 and in the fall of 2007 or had more than 10 jeribs of land owned by the household in the fall of 2007 or had more than 20 poeple in the houdcount in fall of 2007 or a person that has number of people in a household greater than 20 and are observed in summer of 2008. Or if number of jerribs of land counted is more than 10 and are observed in summer of 2008. Lastly if the number of sheep and goats is above 50 and they are observed in summer of 2008.

##Table 4 Question 5: Both of our coefficient estimates were very similar as the treatment for girls was off by 0.004 and the treatment for men is 0.001. The standard error for women was exactly the same while the standard error for men is unknown from table four.

##Table 4 Question 6: The coefficients on all estimates are minimally different with the largest change being the difference between the treated groups in column 2 and 4, (the difference is .018). The standard errors are also negligibly different.

##Table 4 Column 2 Question 6: Like in table 4, both treatments are statically significant. For girls, the only statistically significant control is the child's age in the fall of 2007 (5%). For The males there were more statistically significant controls. The age of the child, (1% level) if the child's family speaks Tajik, (5% level) if the head of household is a farmer or not, (10% level) and lastly the nearest schools distance, (1% level). For both genders the child's age was significant and positive which means that as the child grows older, they are more likely to get a better test score.

##Further Analysis: To start looking at the relationship at the between school we regressed formal enrolled on the distance to the nearest school, whether the child was a girl, an interaction between the two and a variables for distance squared to control for any quadratic patterns. Surprisingly the only statisally significant variable in that regression was whether or not the child was a girl. This means that being farther away does not affect formal enrollment and distance does not affect girls more than boys. The regression output does implicate that women do attend school less than their boy counterparts. Now when re-looking at formal enrollment based on four distance types, short medium short, medium long, and long, we can see that for girl, the amount of distance does not change the fact that treatment is significantly increasing enrollment

or test scores. More specifically as distance increases, the effect of the treatment on tests for girls decreases while it increases for men. Meaning that the treatment is most effective for girls who live close and boys who live farther away. Interestingly the treatment for both boys and girls are mostly similar except when the schools were a medium long distance away meaning that distance does not play a significant role in how the treatment will affect the sexes. Overall it seems as though distance is a more important variable when looking at the test scores for boys and girls and not as much for their attendance rates.