Final Study Data Analysis

April Kim, Jennifer Podracky, Saurav Datta

```
library(ggplot2)
library(data.table)
library(lmtest)
## Loading required package: zoo
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
library(pwr)
library(lsr)
library(cobalt)
library(stringr)
library(AER)
## Loading required package: car
## Loading required package: carData
## Loading required package: sandwich
## Loading required package: survival
```

Read in data and reformat

```
d2 <- d2[, .(userId = UserId,</pre>
             treatment_seq = as.integer(Treatment.Seq),
             day1_treatment = as.integer(as.character(factor(Q6, levels = c('Not applicable', 'In person
                                                                             'Through digital means'),
                                                             labels = c(0, 2, 1))),
             day2_treatment = as.integer(as.character(factor(Q10, levels = c('Not applicable', 'In pers
                                                                              'Through digital means',
                                                                              'Both in person and throug
                                                              labels = c(0, 2, 1, 3))),
             day3_treatment = as.integer(as.character(factor(Q14, levels = c('Not applicable', 'In pers
                                                                              'Through digital means',
                                                                              'Both in person and throug
                                                              labels = c(0, 2, 1, 3))),
             day1_steps = as.numeric(gsub("\\,", "", Q7)),
             day2_steps = as.numeric(gsub("\\,", "", Q11)),
             day3_steps = as.numeric(gsub("\\,", "", Q15)),
             age_range = as.integer(as.character(factor(Age, levels = c('18 - 24',
                                                                         "25 - 34",
                                                                         "35 - 44",
                                                                         "45 - 54",
                                                                         "55 - 64".
                                                                         "65+"),
                                                        labels = c(0, 1, 2, 3, 4, 5)))),
             # gender = factor(Gender),
             gender = as.integer(as.character(factor(Gender, levels = c('Male', 'Female', 'Gender non-c
                                                     labels = c(0, 1, 2))),
             lives_with_others = as.integer(as.character(factor(Living.Situation, levels = c('Alone', ''
                                                                 labels = c(0, 1))),
             # know_us = factor(Q17),
             know_us = as.integer(as.character(factor(Q17, levels = c('No', 'Yes'),
                                                      labels = c(0, 1))),
             location_lat = as.double(LocationLatitude),
             location_long = as.double(LocationLongitude)
)]
## Warning in eval(jsub, SDenv, parent.frame()): NAs introduced by coercion
## Warning in eval(jsub, SDenv, parent.frame()): NAs introduced by coercion
## Warning in eval(jsub, SDenv, parent.frame()): NAs introduced by coercion
d2$gender[is.na(d2$gender)] <- 2
d2$age range[is.na(d2$age range)] <- 6
d2$lives_with_others[is.na(d2$lives_with_others)] <- 2
d2$know_us[is.na(d2$know_us)] <- 2
head(d2, 5)
      userId treatment_seq day1_treatment day2_treatment day3_treatment
##
## 1:
          82
                         6
## 2:
          57
                         3
                                        1
                                                       0
                                                                       2
## 3:
          89
                         4
                                       NA
                                                      NA
                                                                      NA
                         3
                                                       0
                                                                       2
## 4:
          69
                                        1
```

```
## 5:
                        3
                                                      0
                                       1
     day1_steps day2_steps day3_steps age_range gender lives_with_others
                     5040
             NA
                                 3788
                                            1
                     13959
## 2:
          21290
                                13717
                                              0
                                                     0
                                                                       1
## 3:
             NA
                        NA
                                   NA
                                              1
                                                     0
## 4:
           6343
                      3247
                                10198
                                                     0
                                              1
                                                                       1
          13624
                      5406
                                 7851
##
     know_us location_lat location_long
## 1:
          1
                 41.89250
                               -87.7895
                 37.75101
## 2:
           1
                               -97.8220
## 3:
           1
                 37.97240
                              -122.3369
## 4:
                 40.37070
                               -74.0084
           1
## 5:
           1
                 42,41730
                               -71.1087
#Covariate Balance Check
bal.tab(treatment_seq ~ gender + age_range + lives_with_others + know_us + location_lat + location_long
       data = d2
## Balance Measures
##
                       Type Corr.Un
                    Contin. -0.0625
## gender
## age_range
                    Contin. -0.0099
## lives_with_others Contin. -0.0105
## know_us
                    Contin. 0.0214
                    Contin. 0.0157
## location_lat
                    Contin. -0.0480
## location_long
##
## Sample sizes
##
      Total
## All
cov_check <- lm(treatment_seq ~ gender + age_range + lives_with_others + know_us + location_lat + locat</pre>
               data = d2)
summary(cov_check)
##
## lm(formula = treatment_seq ~ gender + age_range + lives_with_others +
      know_us + location_lat + location_long, data = d2)
##
##
## Residuals:
##
      Min
               1Q Median
                               3Q
## -2.7912 -1.4863 0.1883 1.4598 2.6363
##
## Coefficients:
                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                     1.1350717 4.1690323
                                           0.272
                                                     0.786
## gender
                    -0.1639576 0.3999806 -0.410
                                                     0.683
                    -0.0003737 0.1750913 -0.002
## age_range
                                                     0.998
## lives_with_others -0.1171429 0.7099934 -0.165
                                                     0.869
## know_us 0.0212714 0.3840969 0.055
                                                     0.956
## location_lat
                    0.0465410 0.0841290 0.553
                                                     0.582
## location_long -0.0082195 0.0124256 -0.661
                                                     0.511
```

```
##
## Residual standard error: 1.784 on 68 degrees of freedom
## Multiple R-squared: 0.01086, Adjusted R-squared: -0.07641
## F-statistic: 0.1245 on 6 and 68 DF, p-value: 0.993
```

Checking for ordering/priming effect

Is previous day's treatment highly predictive of how many steps are taken today?

```
# n = 75
df <- d2
# remove subjects/rows who were non-compliant
# n = 24
df <- df[rowSums(is.na(df[,c(3:8)])) != ncol(df[,c(3:8)]), ]</pre>
head(df, 5)
##
      userId treatment_seq day1_treatment day2_treatment day3_treatment
## 1:
                                                                        0
          82
                         6
                                         0
                                                         1
## 2:
          57
                         3
                                                                        2
                                         1
                                                         0
                                                                        2
## 3:
          69
                         3
                                         1
                                                         0
## 4:
          85
                         3
                                                         0
                                                                        2
                                         1
                                                         2
## 5:
          66
                         4
##
      day1_steps day2_steps day3_steps age_range gender lives_with_others
## 1:
              NA
                       5040
                                  3788
           21290
## 2:
                      13959
                                  13717
                                                0
                                                       0
                                                                          1
            6343
                       3247
## 3:
                                  10198
                                                1
                                                       0
                                                                          1
## 4:
           13624
                       5406
                                   7851
                                                1
                                                        1
                                                                          1
            7016
                       1211
                                   5717
     know_us location_lat location_long
##
## 1:
          1
                  41.89250
                                -87.7895
## 2:
           1
                  37.75101
                                -97.8220
## 3:
            1
                  40.37070
                                 -74.0084
                  42.41730
                                 -71.1087
## 4:
            1
## 5:
            1
                  42.35760
                                 -71.0514
# day 3 steps using day 1 and 2 treatment
m1 <- lm(day3_steps ~ day1_treatment + day2_treatment, df)</pre>
summary(m1)
##
## Call:
## lm(formula = day3_steps ~ day1_treatment + day2_treatment, data = df)
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
## -7119.9 -2316.1 -462.4 1377.2 8580.4
##
## Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
##
```

```
## (Intercept)
                   7505.7
                               921.5 8.145 2.12e-10 ***
## day1_treatment
                   -385.7
                               672.7 -0.573
                                                0.569
## day2 treatment
                   -365.5
                               600.8 -0.608
                                                0.546
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 3440 on 45 degrees of freedom
     (3 observations deleted due to missingness)
## Multiple R-squared: 0.0142, Adjusted R-squared: -0.02961
## F-statistic: 0.3241 on 2 and 45 DF, p-value: 0.7248
# ATE (standard error)
print(paste0("Estimated effect of day1 treatment: ", signif(m1$coefficients[2], 3),
" (", signif(coef(summary(m1))[2,2], 3), ")"))
## [1] "Estimated effect of day1 treatment: -386 (673)"
print(paste0("Estimated effect of day2 treatment: ", signif(m1$coefficients[3], 3),
" (", signif(coef(summary(m1))[3,2], 3), ")"))
## [1] "Estimated effect of day2 treatment: -366 (601)"
# include days1,2 steps as covariates to uhderstand
# subjects' step counts hange as a function of
# treatment against want they would typically do
m2 <- lm(day3_steps ~ day1_treatment + day2_treatment + day1_steps + day2_steps, df)
summary(m2)
##
## Call:
## lm(formula = day3_steps ~ day1_treatment + day2_treatment + day1_steps +
      day2_steps, data = df)
##
## Residuals:
##
      Min
                               3Q
               1Q Median
                                      Max
## -8183.6 -1375.5
                     61.2 1536.6 5057.0
##
## Coefficients:
                  Estimate Std. Error t value Pr(>|t|)
##
                                       2.306 0.02614 *
## (Intercept)
                 2602.4071 1128.7513
                            517.7869 -0.936 0.35444
## day1_treatment -484.8396
## day2_treatment -296.7856
                             460.9784 -0.644 0.52319
                                       1.985 0.05373 .
## day1_steps
                    0.2442
                               0.1230
## day2_steps
                    0.4542
                               0.1341
                                       3.386 0.00155 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 2623 on 42 degrees of freedom
     (4 observations deleted due to missingness)
## Multiple R-squared: 0.455, Adjusted R-squared: 0.4031
## F-statistic: 8.766 on 4 and 42 DF, p-value: 3.076e-05
```

```
## [1] "Estimated effect of day2 treatment: -297 (461)"
```

We do not see that the previous days' treatment assignments to predict the last day's step count is highgly predictive and significant, which is super for us!

Condense treatment sequence to 1 treatment

```
df1 <- df[,-c(4,5,7,8)]
df2 <- df[,-c(3,5,6,8)]
df3 <- df[,-c(3,4,6,7)]
names(df1)[names(df1) == "day1_treatment"] = "treatment"
names(df1)[names(df1) == "day1_steps"] = "steps"
names(df2)[names(df2) == "day2_treatment"] = "treatment"
names(df2)[names(df2) == "day2_steps"] = "steps"
names(df3)[names(df3) == "day3_treatment"] = "treatment"
names(df3)[names(df3) == "day3_steps"] = "steps"
d <- rbind(df1, df2, df3)
# combine digital and in person treatment as one
d$treatment2 <- ifelse(d$treatment == 0, 0, 1)
d$outcome <- ifelse(d$steps > 5000, 1, 0)
```

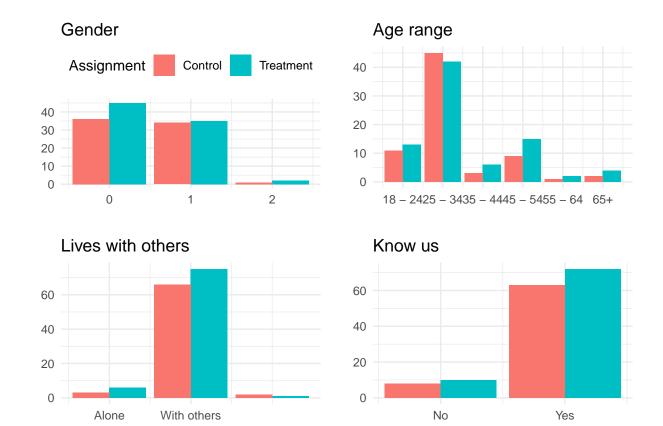
```
##
      userId treatment_seq treatment steps age_range gender lives_with_others
## 1:
                         6
                                    0
                                         NA
                                                    1
                                                            0
## 2:
          57
                         3
                                    1 21290
                                                    0
                                                            0
                                                                              1
          69
                         3
                                    1 6343
                                                            0
                                                                              1
## 3:
                                                    1
          85
                         3
## 4:
                                    1 13624
                                                    1
                                                                              1
                                                            1
                                    1 7016
## 5:
          66
                         4
                                                                              1
##
      know_us location_lat location_long treatment2 outcome
## 1:
           1
                  41.89250
                                -87.7895
                                                   0
                                                           NA
## 2:
            1
                  37.75101
                                 -97.8220
                                                   1
                                                            1
                  40.37070
## 3:
            1
                                 -74.0084
                                                   1
                                                            1
## 4:
            1
                  42.41730
                                 -71.1087
                                                   1
                                                            1
## 5:
            1
                  42.35760
                                 -71.0514
                                                   1
                                                            1
```

Check covariates again after transforming and create plots (Maybe we can just check covariates here?)

```
#Covariate Balance Check
bal.tab(treatment2 ~ gender + age_range + lives_with_others + know_us + location_lat + location_long,
       data = d
## Note: estimand and s.d.denom not specified; assuming ATE and pooled.
## Balance Measures
##
                       Type Diff.Un
## gender
                    Contin. -0.0582
## age_range
                   Contin. 0.2103
## lives_with_others Contin. -0.1690
                    Binary -0.0093
Contin. 0.0444
## know_us
## location lat
                    Contin. 0.0661
## location_long
## Sample sizes
##
      Control Treated
         71
## All
                   82
cov_check1 <- lm(treatment_seq ~ gender + age_range + lives_with_others + know_us + location_lat + locat</pre>
               data = d
summary(cov_check1)
##
## Call:
## lm(formula = treatment_seq ~ gender + age_range + lives_with_others +
      know_us + location_lat + location_long, data = d)
##
##
## Residuals:
##
       Min
                 1Q Median
                                  3Q
## -2.45660 -1.45557 -0.07703 1.47990 2.76916
##
## Coefficients:
##
                    Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                    -0.949777
                               3.094677 -0.307
                                                0.759
                                         0.809
                                                 0.420
## gender
                    0.222693 0.275319
## age_range -0.037168
                               0.124112 -0.299 0.765
                                         0.430
## lives_with_others 0.220037
                                                  0.668
                               0.511785
## know_us
                    0.316655
                               0.508372
                                         0.623
                                                  0.534
## location_lat
                   0.074428
                               0.056090
                                         1.327
                                                  0.187
                   -0.007474
                               0.009532 -0.784
## location_long
                                                   0.434
## Residual standard error: 1.701 on 146 degrees of freedom
## Multiple R-squared: 0.02282,
                                  Adjusted R-squared:
## F-statistic: 0.5683 on 6 and 146 DF, p-value: 0.755
require(gridExtra)
```

Loading required package: gridExtra

```
d.gender <- d[, c("gender", "treatment2")]</pre>
p_gender <- ggplot(d.gender, aes(x=gender, fill = factor(treatment2))) +</pre>
  geom_bar(stat="count", position=position_dodge()) +
  theme_minimal() + theme(legend.position="top") +
  xlab("") + ylab("") + ggtitle("Gender") +
  guides(fill = guide_legend(title = "Assignment")) +
  scale_fill_discrete(labels = c("Control", "Treatment"))
d.age <- d[, c("age_range", "treatment2")]</pre>
p_age <- ggplot(d.age, aes(x=age_range, fill = factor(treatment2))) +</pre>
  geom_bar(stat="count", position=position_dodge()) +
  theme_minimal() + theme(legend.position="none") +
  xlab("") + ylab("") + ggtitle("Age range") +
  # quides(fill = quide_legend(title = "Assignment")) +
  # scale_fill_discrete(labels = c("Control", "Treatment")) +
  scale_x_continuous(breaks = c(0, 1, 2, 3, 4, 5),
                   labels = c('18 - 24')
                               "25 - 34",
                               "35 - 44",
                               "45 - 54".
                               "55 - 64",
                               "65+"))
d.others <- d[, c("lives_with_others", "treatment2")]</pre>
p others <- ggplot(d.others, aes(x=lives with others, fill = factor(treatment2))) +</pre>
  geom_bar(stat="count", position=position_dodge()) +
  theme minimal() + theme(legend.position="none") +
  xlab("") + ylab("") + ggtitle("Lives with others") +
  # guides(fill = guide_legend(title = "Assignment")) +
  # scale_fill_discrete(labels = c("Control", "Treatment")) +
  scale_x_continuous(breaks = c(0, 1),
                     labels = c('Alone', 'With others'))
d.know_us <- d[, c("know_us", "treatment2")]</pre>
p_know_us <- ggplot(d.know_us, aes(x=know_us, fill = factor(treatment2))) +</pre>
  geom_bar(stat="count", position=position_dodge()) +
  theme_minimal() + theme(legend.position="none") +
  xlab("") + ylab("") + ggtitle("Know us") +
  # quides(fill = quide_legend(title = "Assignment")) +
  # scale_fill_discrete(labels = c("Control", "Treatment")) +
  scale_x_continuous(breaks = c(0, 1),
                     labels = c('No', 'Yes'))
grid.arrange(p_gender, p_age, p_others, p_know_us,
             ncol = 2)
```



T-test and power calculations

```
### Control vs treatment (digital+in person)
t.test(d[treatment == 0]$outcome, d[treatment == 1]$outcome)
##
##
    Welch Two Sample t-test
##
## data: d[treatment == 0]$outcome and d[treatment == 1]$outcome
## t = -1.5548, df = 87.831, p-value = 0.1236
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
  -0.29649330 0.03620224
## sample estimates:
## mean of x mean of y
## 0.7076923 0.8378378
# since we fail to reject the null hypothesis,
# let's calculate number of subjects needed for 80% power
effect_size <- cohensD(d[treatment == 0] $outcome, d[treatment == 1] $outcome)
pwr.t.test(power = 0.8, d = effect_size, sig.level = 0.05, type = "two.sample")
```

```
## Two-sample t test power calculation
##
## n = 172.172
## d = 0.3028018
## sig.level = 0.05
## power = 0.8
## alternative = two.sided
##
## NOTE: n is number in *each* group
```

data manipulation for paired t test!

```
'%!in%' <- function(x,y)!('%in%'(x,y))
# limit to those who complied and
# get data frame in workable format
d_complied <- rbindlist(list(subset(df, treatment_seq == 1 & df$day1_treatment == assigned_treatment_se</pre>
                                     & df$day2_treatment == assigned_treatment_seq[1,2]
                                    & df$day3_treatment == assigned_treatment_seq[1,3]),
                              subset(df, treatment_seq == 2 & df$day1_treatment == assigned_treatment_se
                                    & df$day2_treatment == assigned_treatment_seq[2,2]
                                     & df$day3_treatment == assigned_treatment_seq[2,3]),
                              subset(df, treatment_seq == 3 & df$day1_treatment == assigned_treatment_se
                                     & df$day2_treatment == assigned_treatment_seq[3,2]
                                     & df$day3_treatment == assigned_treatment_seq[3,3]),
                              subset(df, treatment_seq == 4 & df$day1_treatment == assigned_treatment_se
                                     & df$day2_treatment == assigned_treatment_seq[4,2]
                                     & df$day3_treatment == assigned_treatment_seq[4,3]),
                              subset(df, treatment_seq == 5 & df$day1_treatment == assigned_treatment_se
                                     & df$day2_treatment == assigned_treatment_seq[5,2]
                                     & df$day3_treatment == assigned_treatment_seq[5,3]),
                              subset(df, treatment_seq == 6 & df$day1_treatment == assigned_treatment_se
                                     & df$day2_treatment == assigned_treatment_seq[6,2]
                                     & df$day3_treatment == assigned_treatment_seq[6,3])
))
d_not_complied <- subset(df, userId %!in% d_complied$userId)</pre>
d not complied ok <- subset(d not complied, d not complied$day1 treatment != d not complied$day2 treatm
                              d_not_complied$day1_treatment != d_not_complied$day3_treatment &
                              d_not_complied$day2_treatment != d_not_complied$day3_treatment)
d_usable <- rbind(d_complied, d_not_complied_ok)</pre>
d <- rbindlist(list(d_usable[1:nrow(d_usable),c("day1_treatment","day1_steps")],</pre>
                    d_usable[1:nrow(d_usable),c("day2_treatment","day2_steps")],
                    d_usable[1:nrow(d_usable),c("day3_treatment","day3_steps")]))
names(d) <- c("treatment", "steps")</pre>
# interpolate using median steps
d$steps[is.na(d$steps)] <- median(d[which(!is.na(d$steps))]$steps)
d$outcome <- ifelse(d$steps > 5000, 1, 0)
# CACE
ITT_lm <- lm(outcome ~ treatment, data = d)</pre>
summary(ITT_lm)$coefficients[2]
```

[1] -0.04411765

```
# COMMENTED LINES BELOW HAS TO BE FIXED!
# SKIP DOWN TO T TEST IN LINE 290
# combine digital and in person as one treatment as
# "Both in person and through digital means"
# d1 <- d_usable
# d1a <- subset(d1, d1$day1 treatment == 1 & d1$day2 treatment == 2)
# d1a[, treatment_total_steps := day1_steps + day2_steps]
# d1b <- subset(d1, d1$day2_treatment == 1 & d1$day3_treatment == 2)
# d1b[, treatment_total_steps := day2_steps + day3_steps]
# d1c <- subset(d1, d1$day1 treatment == 1 & d1$day3 treatment == 2)
# d1c[, treatment_total_steps := day1_steps + day2_steps]
# d1d <- subset(d1, d1$day2_treatment == 2 & d1$day3_treatment == 1)
# d1d[, treatment_total_steps := day2_steps + day3_steps]
# d1e \leftarrow subset(d1, d1$day1_treatment == 2 & d1$day3_treatment == 1)
# d1e[, treatment_total_steps := day1_steps + day2_steps]
\# d1f \leftarrow subset(d1, d1\$day1\_treatment == 2 \& d1\$day2\_treatment == 1)
# d1f[, treatment_total_steps := day2_steps + day3_steps]
# d1_comb <- rbind(d1a, d1b, d1c, d1d, d1e, d1f)
# d1 <- rbindlist(list(d1_comb[1:nrow(d1_comb),c("day1_treatment","day1_steps")],
                       d1_comb[1:nrow(d1_comb),c("day2_treatment", "day2_steps")],
#
                       d1\_comb[1:nrow(d1\_comb),c("day3\_treatment","day3\_steps")]))
# names(d1) <- c("treatment", "steps")</pre>
# d1[d1$treatment == 1 | d1$treatment == 2, ]$treatment <- 3
# # linear interpolation
# # d1 \leftarrow na.approx(d1, maxqap=5)
# d1 <- data.table(d1)
# d1$steps[is.na(d1$steps)] <- median(d1[which(!is.na(d1$steps))]$steps)</pre>
# d1$outcome <- ifelse(d1$steps > 5000, 1, 0)
t.test(d[treatment == 0]$outcome, d[treatment == 1]$outcome, paired = T)
##
##
   Paired t-test
##
## data: d[treatment == 0]$outcome and d[treatment == 1]$outcome
## t = 0, df = 33, p-value = 1
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1920722 0.1920722
## sample estimates:
## mean of the differences
##
cohensD(d[treatment == 0]$outcome, d[treatment == 1]$outcome, method = "paired")
## [1] 0
```

```
### Control vs in person
t.test(d[treatment == 0]$outcome, d[treatment == 2]$outcome, paired = T)
##
## Paired t-test
## data: d[treatment == 0]$outcome and d[treatment == 2]$outcome
## t = 0.90205, df = 33, p-value = 0.3736
\#\# alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1107732 0.2872437
## sample estimates:
## mean of the differences
##
                0.08823529
cohensD(d[treatment == 0]$outcome, d[treatment == 2]$outcome, method = "paired")
## [1] 0.1547007
### Control vs in person + digital
# t.test(d1[treatment == 0]$outcome, d1[treatment == 3]$outcome)
```