

Final Study Data Analysis

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```
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

```
assigned_treatment_seq <- data.frame(seq_id = c(1,2,3,4,5,6),
                                     day1 = c(0,0,1,1,2,2),
                                     day2 = c(1,2,0,2,0,1),
                                     day3 = c(2,1,2,0,1,0))
d2 <- fread("241 Participant List - Final Study Results - 20181215.csv", na.strings=c("", "NA"))
d2[UserId == 65,]$Q10 <- "In person"
d2[UserId == 13,]$Q6 <- "Through digital means"
# stringsAsFactors = F)
names(d2) <- str_replace_all(names(d2), c(" " = "." , "," = "" ))

# Not applicable = 0
# Through digital means = 1
# In person = 2
# Both in person and through digital means = 3
```

```

d2 <- d2[, .(userId = UserId,
  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 person',
    'Through digital means',
    'Both in person and through',
    labels = c(0, 2, 1, 3)))),
  day3_treatment = as.integer(as.character(factor(Q14, levels = c('Not applicable', 'In person',
    'Through digital means',
    'Both in person and through',
    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-conforming'),
    labels = c(0, 1, 2)))),
  lives_with_others = as.integer(as.character(factor(Living.Situation, levels = c('Alone', 'With others'),
    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, SEnv, parent.frame()): NAs introduced by coercion
```

```
## Warning in eval(jsub, SEnv, parent.frame()): NAs introduced by coercion
```

```
## Warning in eval(jsub, SEnv, 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             0             1             0
## 2:      57             3             1             0             2
## 3:      89             4             NA             NA             NA
## 4:      69             3             1             0             2

```

```
## 5:      85          3          1          0          2
##   day1_steps day2_steps day3_steps age_range gender lives_with_others
## 1:      NA      5040      3788         1         0             1
## 2:    21290    13959    13717         0         0             1
## 3:      NA       NA       NA         1         0             1
## 4:     6343     3247     10198         1         0             1
## 5:    13624     5406     7851         1         1             1
##   know_us location_lat location_long
## 1:      1     41.89250     -87.7895
## 2:      1     37.75101     -97.8220
## 3:      1     37.97240    -122.3369
## 4:      1     40.37070     -74.0084
## 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
## gender        Contin. -0.0625
## age_range      Contin. -0.0099
## lives_with_others Contin. -0.0105
## know_us        Contin.  0.0214
## location_lat   Contin.  0.0157
## location_long  Contin. -0.0480
##
## Sample sizes
##      Total
## All      75
```

```
cov_check <- lm(treatment_seq ~ gender + age_range + lives_with_others + know_us + location_lat + location_long,
                 data = d2)
summary(cov_check)
```

```
##
## Call:
## lm(formula = treatment_seq ~ gender + age_range + lives_with_others +
##      know_us + location_lat + location_long, data = d2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -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
## age_range      -0.0003737  0.1750913  -0.002   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)]), ]

head(df, 5)
```

```
##      userId treatment_seq day1_treatment day2_treatment day3_treatment
## 1:      82             6             0             1             0
## 2:      57             3             1             0             2
## 3:      69             3             1             0             2
## 4:      85             3             1             0             2
## 5:      66             4             1             2             0
##      day1_steps day2_steps day3_steps age_range gender lives_with_others
## 1:      NA      5040      3788          1      0             1
## 2:     21290     13959     13717          0      0             1
## 3:      6343      3247     10198          1      0             1
## 4:     13624      5406      7851          1      1             1
## 5:      7016      1211      5717          0      0             1
##      know_us location_lat location_long
## 1:      1      41.89250      -87.7895
## 2:      1      37.75101      -97.8220
## 3:      1      40.37070      -74.0084
## 4:      1      42.41730      -71.1087
## 5:      1      42.35760      -71.0514
```

```
# day 3 steps using day 1 and 2 treatment
m1 <- lm(day3_steps ~ day1_treatment + day2_treatment, df)
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 '**' 0.01 '*' 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 uherstand
# subjects' step counts hange as a function of
# treatment against waht 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       1Q   Median       3Q      Max
## -8183.6 -1375.5    61.2  1536.6  5057.0
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  2602.4071  1128.7513   2.306  0.02614 *
## day1_treatment -484.8396   517.7869  -0.936  0.35444
## day2_treatment -296.7856   460.9784  -0.644  0.52319
## day1_steps      0.2442    0.1230   1.985  0.05373 .
## day2_steps      0.4542    0.1341   3.386  0.00155 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 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
```

```
print(paste0("Estimated effect of day1 treatment: ", signif(m2$coefficients[2], 3),
            " (", signif(coef(summary(m2))[2,2], 3), ")"))
```

```
## [1] "Estimated effect of day1 treatment: -485 (518)"
```

```
print(paste0("Estimated effect of day2 treatment: ", signif(m2$coefficients[3], 3),
            " (", signif(coef(summary(m2))[3,2], 3), ")"))
```

```
## [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 highly 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)
head(d, 5)
```

```
##      userId treatment_seq treatment steps age_range gender lives_with_others
## 1:      82             6         0   NA          1      0              1
## 2:      57             3         1 21290          0      0              1
## 3:      69             3         1  6343          1      0              1
## 4:      85             3         1 13624          1      1              1
## 5:      66             4         1  7016          0      0              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
## 3:         1    40.37070    -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  
## know_us     Binary -0.0093  
## location_lat Contin.  0.0444  
## location_long Contin.  0.0661  
##  
## Sample sizes  
##      Control Treated  
## All      71      82
```

```
cov_check1 <- lm(treatment_seq ~ gender + age_range + lives_with_others + know_us + location_lat + location_long,  
                 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      Max   
## -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      
## gender         0.222693    0.275319   0.809    0.420      
## age_range     -0.037168    0.124112  -0.299    0.765      
## lives_with_others 0.220037    0.511785   0.430    0.668      
## know_us        0.316655    0.508372   0.623    0.534      
## location_lat    0.074428    0.056090   1.327    0.187      
## location_long  -0.007474    0.009532  -0.784    0.434      
##  
## Residual standard error: 1.701 on 146 degrees of freedom  
## Multiple R-squared:  0.02282,    Adjusted R-squared:  -0.01733   
## 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")]
p_gender <- ggplot(d.gender, aes(x=gender, fill = factor(treatment2))) +
  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")) +
  scale_x_continuous(breaks = c(0, 1, 2),
    labels = c('Male', 'Female', 'Gender non-conforming'))

d.age <- d[, c("age_range", "treatment2")]
p_age <- ggplot(d.age, aes(x=age_range, fill = factor(treatment2))) +
  geom_bar(stat="count", position=position_dodge()) +
  theme_minimal() + theme(legend.position="none") +
  xlab("") + ylab("") + ggtitle("Age range") +
  # guides(fill = guide_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+'))

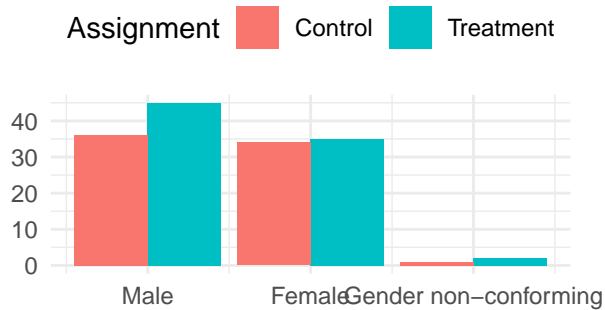
d.others <- d[, c("lives_with_others", "treatment2")]
p_others <- ggplot(d.others, aes(x=lives_with_others, fill = factor(treatment2))) +
  geom_bar(stat="count", position=position_dodge()) +
  theme_minimal() + theme(legend.position="none") +
  xlab("") + ylab("") + ggtitle("Lives with others") +
  # 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")]
p_know_us <- ggplot(d.know_us, aes(x=know_us, fill = factor(treatment2))) +
  geom_bar(stat="count", position=position_dodge()) +
  theme_minimal() + theme(legend.position="none") +
  xlab("") + ylab("") + ggtitle("Know us") +
  # guides(fill = guide_legend(title = "Assignment")) +
  # scale_fill_discrete(labels = c("Control", "Treatment")) +
  scale_x_continuous(breaks = c(0, 1),
    labels = c('No', 'Yes'))

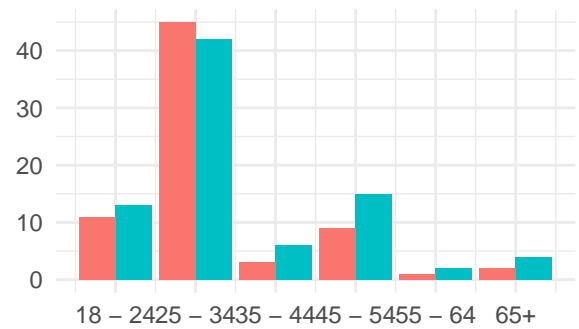
grid.arrange(p_gender, p_age, p_others, p_know_us,
  ncol = 2)

```

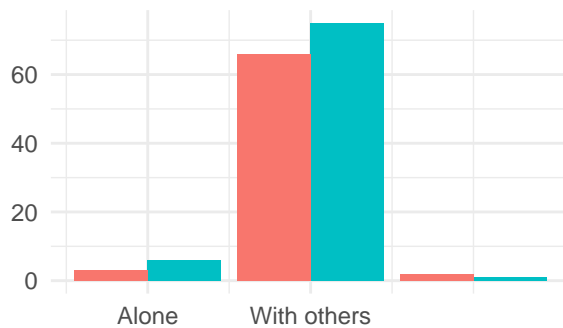

Gender



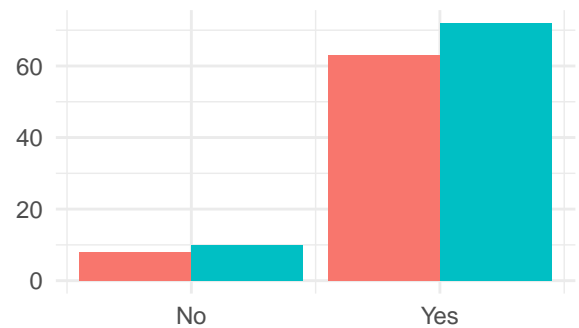
Age range



Lives with others



Know us



T-test and power calculations

```
### Control vs treatment (digital+in person)
t.test(d[treatment2 == 0]$outcome, d[treatment2 == 1]$outcome)
```

```
##
## Welch Two Sample t-test
##
## data: d[treatment2 == 0]$outcome and d[treatment2 == 1]$outcome
## t = -0.91261, df = 131.56, p-value = 0.3631
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.21320328 0.07858789
## sample estimates:
## mean of x mean of y
## 0.7076923 0.7750000
```

```
# since we fail to reject the null hypothesis,
# let's calculate number of subjects needed for 80% power
effect_size <- cohensD(d[treatment2 == 0]$outcome, d[treatment2 == 1]$outcome)
pwr.t.test(power = 0.8, d = effect_size, sig.level = 0.05, type = "two.sample")
```

```
##
```

```
##      Two-sample t test power calculation
##
##          n = 664.7977
##          d = 0.1537758
##      sig.level = 0.05
##          power = 0.8
##      alternative = two.sided
##
## NOTE: n is number in each group
```

```
# if we were to look at digital and in person separately
# digital
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
```

```
# in person
t.test(d[treatment == 0]$outcome, d[treatment == 2]$outcome)
```

```
##
## Welch Two Sample t-test
##
## data: d[treatment == 0]$outcome and d[treatment == 2]$outcome
## t = -0.1478, df = 90.702, p-value = 0.8828
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.1911623 0.1646864
## sample estimates:
## mean of x mean of y
## 0.7076923 0.7209302
```

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$day1_treatment == assigned_treatment_seq
                                & df$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_seq
                                & 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_seq[3,2]
        & df$day2_treatment == assigned_treatment_seq[3,2]
        & df$day3_treatment == assigned_treatment_seq[3,3]),
subset(df, treatment_seq == 4 & df$day1_treatment == assigned_treatment_seq[4,2]
        & df$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_seq[5,2]
        & df$day2_treatment == assigned_treatment_seq[5,2]
        & df$day3_treatment == assigned_treatment_seq[5,3]),
subset(df, treatment_seq == 6 & df$day1_treatment == assigned_treatment_seq[6,2]
        & df$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)
d_not_complied_ok <- subset(d_not_complied, d_not_complied$day1_treatment != d_not_complied$day2_treatment
        & 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)
d <- rbindlist(list(d_usable[1:nrow(d_usable),c("day1_treatment", "day1_steps")],
        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")
# 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)
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 <- subset(d1, d1$day1_treatment == 2 & d1$day3_treatment == 1)
# d1e[, treatment_total_steps := day1_steps + day2_steps]
# d1f <- 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")
# d1[d1$treatment == 1 | d1$treatment == 2, ]$treatment <- 3
# # linear interpolation
# # d1 <- na.approx(d1, maxgap=5)
# d1 <- data.table(d1)
# d1$steps[is.na(d1$steps)] <- median(d1[which(!is.na(d1$steps))]$steps)
# d1$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
## 0
```

```
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)
```