

# IPTW Causal Project

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.0 --
```

```
## v ggplot2 3.3.3    v purrr   0.3.4
## v tibble  3.1.0    v dplyr   1.0.5
## v tidyr   1.1.3    v stringr 1.4.0
## v readr   1.4.0    v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
library(readxl)
```

IPTW estimand:

$$\Psi(\mathbb{P}_0) = \mathbb{E}_0\left[\frac{\mathbb{I}(A=1)}{\mathbb{P}_\mu(A=1|W)}Y\right] - \mathbb{E}_0\left[\frac{\mathbb{I}(A=0)}{\mathbb{P}_\mu(A=0|W)}Y\right]$$

```
ObsData <- read.csv("slpexcov1517.csv")
```

```
ObsData <- ObsData %>% dplyr::select(-SEQN, -exminwk, -slphrs, -household, -income, -snoring,
                                     -apnea, -bmecat, -smoke, -alcohol, -phq9)
```

```
ObsData <- ObsData %>% mutate(A = targetex) %>% mutate(Y = targetslp) %>%
  dplyr::select(-targetex, -targetslp)
```

```
ObsData <- na.omit(ObsData)
```

```
names(ObsData)
```

```
## [1] "age"      "raceeth"  "educ"     "marital"  "bmi"      "waist"
## [7] "depressed" "A"        "Y"
```

```
summary(ObsData)
```

```
##      age      raceeth      educ      marital
## Min.   :20.00  Min.    :1.000  Min.    :1.000  Min.    :1.000
## 1st Qu.:31.00  1st Qu.:1.000  1st Qu.:2.000  1st Qu.:1.000
## Median :43.00  Median :2.000  Median :3.000  Median :2.000
## Mean   :42.94  Mean     :2.299  Mean     :2.574  Mean     :1.644
## 3rd Qu.:55.00  3rd Qu.:3.000  3rd Qu.:3.000  3rd Qu.:2.000
## Max.   :64.00  Max.     :4.000  Max.     :4.000  Max.     :2.000
```

```
##      bmi      waist      depressed      A
## Min.   :15.50   Min.    : 62.30   Min.    :0.00000   Min.    :0.0000
## 1st Qu.:24.82   1st Qu.: 89.83   1st Qu.:0.00000   1st Qu.:0.0000
## Median :28.35   Median : 99.30   Median :0.00000   Median :0.0000
## Mean   :29.32   Mean    :101.28   Mean    :0.07017   Mean    :0.4163
## 3rd Qu.:32.80   3rd Qu.:110.60   3rd Qu.:0.00000   3rd Qu.:1.0000
## Max.   :61.90   Max.    :169.60   Max.    :1.00000   Max.    :1.0000
##      Y
## Min.   :0.0000
## 1st Qu.:1.0000
## Median :1.0000
## Mean   :0.7783
## 3rd Qu.:1.0000
## Max.   :1.0000
```

## 1) Create the propensity scores

First fit the logistic regression model:

```
fit<-glm(A~factor(raceeth)+factor(educ)+factor(marital)+bmi+waist+factor(depressed),
        family="binomial",data=ObsData)
```

Get propensity scores:

```
prob.1W <- predict(fit, type= "response") #prediced probability of getting the exercise
prob.0W <- 1 - prob.1W #prediced probability of not getting the exercise
```

look at distribution of propensity scores:

```
summary(prob.1W)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.04543 0.28218 0.40203 0.41632 0.54587 0.86707
```

```
summary(prob.0W)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.1329 0.4541 0.5980 0.5837 0.7178 0.9546
```

## 2) Create the weights:

```
wt1 <- as.numeric(ObsData$A==1)/prob.1W
wt0 <- as.numeric(ObsData$A==0)/prob.0W
```

Look at weights:

```
summary(wt1)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  0.000   0.000   0.000   1.013   1.844   22.014
```

```
summary(wt0)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  0.0000   0.0000   1.2414   0.9942   1.6147   6.7218
```

### 3) Estimate:

```
IPTW<- mean( wt1*ObsData$Y) - mean( wt0*ObsData$Y)
```

```
IPTW
```

```
## [1] 0.0660337
```

### 4) Arbitrarily truncate weights at 10, to see what happens:

First see how many weights are greater than 10:

```
sum(wt1>10)
```

```
## [1] 4
```

```
sum(wt0>10)
```

```
## [1] 0
```

```
wt1.trunc<- wt1
wt1.trunc[ wt1.trunc>10] <-10
```

```
wt0.trunc<- wt0
wt0.trunc[ wt0.trunc>10] <-10
```

IPTW with truncated weights at 10:

```
mean(wt1.trunc*ObsData$Y) - mean( wt0.trunc*ObsData$Y)
```

```
## [1] 0.06216204
```

What about truncated at 5?

```
sum(wt1>5)
```

```
## [1] 60
```

```
sum(wt0>5)
```

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

```
wt1.trunc5<- wt1  
wt1.trunc5[ wt1.trunc5>5] <-5
```

```
wt0.trunc5<- wt0  
wt0.trunc5[ wt0.trunc5>5] <-5
```

IPTW with truncated weights at 10:

```
mean(wt1.trunc5*ObsData$Y) - mean( wt0.trunc5*ObsData$Y)
```

```
## [1] 0.04196059
```

## 5) Stabilized IPTW estimator - Modified Horwitz Thompson estimator

```
mean( wt1*ObsData$Y)/mean( wt1) - mean( wt0*ObsData$Y)/mean( wt0)
```

```
## [1] 0.05142672
```