## P8131-HW8-yz4184

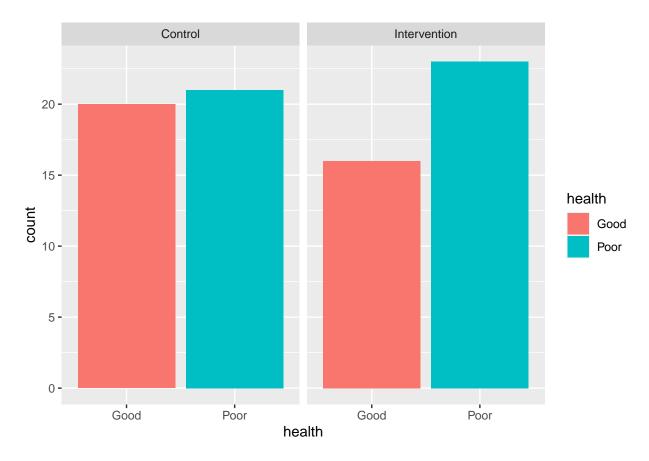
## Yunlin Zhou

```
library(tidyverse)
library(ggplot2)
```

## Part A

First, we plot the count of participants health self-rating counts in different treatment groups.

```
health_df %>%
  filter(time == 1) %>%
  ggplot(aes(x= health, fill = health)) +
  geom_bar() +
  facet_grid(cols = vars(txt))
```



At the time of randomization, the control group doesn't show significant difference in health self-rating. However, the intervention group shows more poor self-ratings than good self-ratings.

Then, we would like to use logistic model to test the relationship between randomized group assignment and participants health self-rating.

```
##
## Call:
## glm(formula = health ~ txt, family = binomial(link = "logit"),
       data = health random)
##
##
## Deviance Residuals:
##
      Min
               1Q
                   Median
                                3Q
                                       Max
##
   -1.335
          -1.198
                     1.028
                             1.157
                                      1.157
##
## Coefficients:
##
                    Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                    0.04879
                                0.31244
                                          0.156
                                                    0.876
```

```
## txtIntervention 0.31412 0.45122 0.696 0.486
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 110.10 on 79 degrees of freedom
## Residual deviance: 109.62 on 78 degrees of freedom
## AIC: 113.62
##
## Number of Fisher Scoring iterations: 4
```

The odds ratio of having good health is expected to be 1.3690476 for intervention group compared to control group.

The p value for the coefficient is 0.486 which is greater than 0.05, which indicates strong evidence for the null hypothesis that there is no correlation between treatment group assignment and health status at the time of randomization.

## Part b