## PS Bootstrap Binary Standardization

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### Generating data with true log odds ratio and its standard deviation

```
pre_data <- defData(varname = "L1", formula = "0", variance = 1,</pre>
                 dist = "normal")
pre_data <- defData(pre_data, varname = "L2", formula = "0", variance = 1,</pre>
                 dist = "normal")
pre_data <- defData(pre_data, varname = "L3", formula = "0", variance = 1,</pre>
                 dist = "normal")
pre_data <- defData(pre_data, varname = "A",</pre>
                     formula = " 0.5*L1 + 0.27*L2 -0.17*L3",
                 dist = "binary", link = "logit")
pre_data <- defData(pre_data, varname = "Y",</pre>
                     formula = "0.5*A + 0.8*L2 + -0.1*L3",
                 dist = "binary", link = "logit")
set.seed(7777)
df <- genData(1000, pre_data)</pre>
expit <- function(beta) {</pre>
    return(exp(beta)/(1 + exp(beta)))
ATE <- expit(sum(0.5 + 0.8*df$L2 - 0.1*df$L3)) - expit(sum(0.8*df$L2 - 0.1*df$L3))
# this is not true ATE
# True log odds ratio: 0.5
```

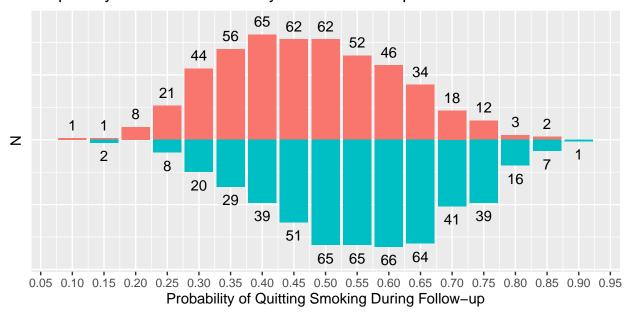
### Propensity Score Model

### 500 pairs Propensity Score distribution

```
df %>%
  mutate(ps.grp = round(ps/0.05) * 0.05) %>%
  group_by(A, ps.grp) %>%
  summarize(n = n()) %>%
  ungroup() %>%
  mutate(n2 = ifelse(A == 0, yes = n, no = -1*n)) %>%
  ggplot(aes(x = ps.grp, y = n2, fill = as.factor(A))) +
```

## 'summarise()' has grouped output by 'A'. You can override using the '.groups'
## argument.

### Propensity Score Distribution by Treatment Group





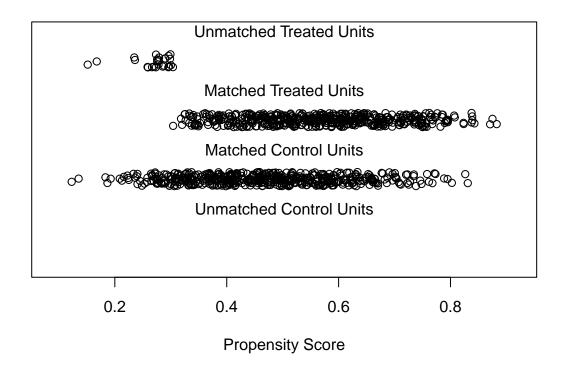
### Nearest neighbor propensity score matching

```
summary(matched)[2]
```

```
## $nn
##
                 Control Treated
## All (ESS)
                     487
                             513
## All
                     487
                             513
                     487
                             487
## Matched (ESS)
## Matched
                     487
                             487
## Unmatched
                              26
## Discarded
                               0
```

```
plot(matched, type = "jitter", interactive = FALSE)
```

## **Distribution of Propensity Scores**

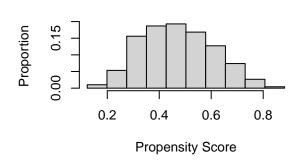


plot(matched, type = "histogram")

#### **Raw Treated Matched Treated** 0.15 Proportion 0.15 Proportion 0.00 0.00 0.2 0.4 0.6 0.8 0.2 0.4 0.6 8.0 Propensity Score **Propensity Score**

# 0.2 0.4 0.6 0.8 Propensity Score

**Raw Control** 



**Matched Control** 

```
matched_df <-
match.data(matched)</pre>
```

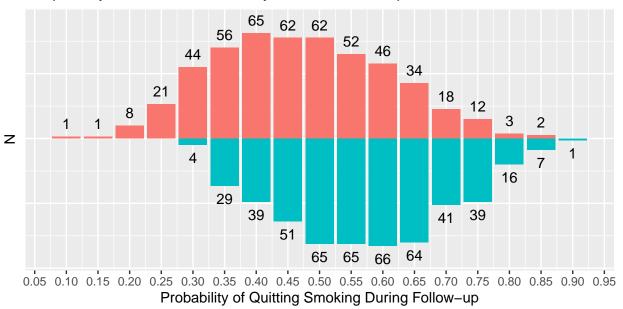
### 495 pairs propensity score distribution

```
matched_df %>%
  mutate(ps.grp = round(ps/0.05) * 0.05) %
  group_by(A, ps.grp) %>%
  summarize(n = n()) \%>\%
  ungroup() %>%
  mutate(n2 = ifelse(A == 0, yes = n, no = -1*n)) \%%
  ggplot(aes(x = ps.grp, y = n2, fill = as.factor(A))) +
  geom_bar(stat = 'identity', position = 'identity') +
  geom_text(aes(label = n, x = ps.grp, y = n2 + ifelse(A == 0, 8, -8))) +
  xlab('Probability of Quitting Smoking During Follow-up') +
  ylab('N') +
  ggtitle('Propensity Score Distribution by Treatment Group') +
  scale_fill_discrete('') +
  scale_x_continuous(breaks = seq(0, 1, 0.05)) +
  theme(legend.position = 'bottom', legend.direction = 'vertical',
        axis.ticks.y = element_blank(),
        axis.text.y = element_blank())
```

## 'summarise()' has grouped output by 'A'. You can override using the '.groups'

### ## argument.

# Propensity Score Distribution by Treatment Group





### simple bootstrap

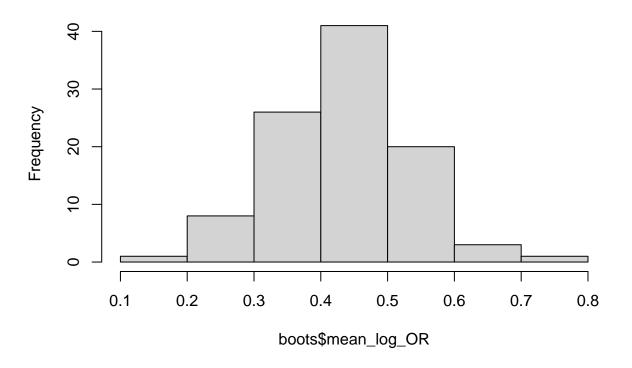
```
nboot <- 100
# set up a matrix to store results
boots <- data.frame(i = 1:nboot,</pre>
                     se_{OR} = NA,
                     mean_log_OR = NA,
                     mean1 = NA,
                     mean0 = NA,
                     difference = NA
# loop to perform the bootstrapping
for(i in 1:nboot) {
  # sample with replacement
  sampl <- matched_df %>% filter(subclass %in% sample(levels(subclass),487, replace = TRUE))
  bootmod \leftarrow glm(Y \sim A + ps, data = sampl,
                  weights = weights, family = binomial)
  # create new data sets
  sampl.treated <- sampl %>%
```

```
mutate(A = 1)
sampl.untreated <- sampl %>%
  mutate(A = 0)
# predict values
sampl.treated$pred.y <-</pre>
  predict(bootmod, sampl.treated, type = "response")
sampl.untreated$pred.y <-</pre>
  predict(bootmod, sampl.untreated, type = "response")
 # output results
boots[i, "mean_log_OR"] <- summary(bootmod)$coeff[2,1]</pre>
boots[i, "se_OR"] <- summary(bootmod)$coeff[2,2]</pre>
boots[i, "mean1"] <- mean(sampl.treated$pred.y)</pre>
boots[i, "mean0"] <- mean(sampl.untreated$pred.y)</pre>
boots[i, "difference"] <- boots[i, "mean1"] - boots[i, "mean0"]</pre>
se_ATE <- sd(boots$difference)</pre>
ATE <- mean(boots$difference)
# once loop is done, print the results
if (i == nboot){
  cat("ATE:")
  cat(ATE)
  cat("\n")
  cat("\n")
  cat("se_ATE:")
  cat(se_ATE)
  cat("\n")
  cat("\n")
  cat("95% CI for ATE:")
  cat(ATE - 1.96*se_ATE,
      ",",
      ATE + 1.96*se_ATE)
  cat("\n")
  cat("\n")
  cat("log OR:")
  cat(mean(boots$mean_log_OR))
  cat("\n")
  cat("\n")
  cat("se_OR:")
  cat(mean(boots$se_OR))
  cat("\n")
  cat("\n")
  cat("95% CI for log odds ratio:")
  cat(mean(boots$mean_log_OR) - 1.96*mean(boots$se_OR),
      ",",
```

```
mean(boots$mean_log_OR) + 1.96*mean(boots$se_OR))
}

## ATE:0.1064552
##
## se_ATE:0.02492412
##
## 95% CI for ATE:0.05760391 , 0.1553065
##
## log OR:0.434012
##
## se_OR:0.1739941
##
## 95% CI for log odds ratio:0.09298361 , 0.7750404
hist(boots$mean_log_OR)
```

# Histogram of boots\$mean\_log\_OR



hist(boots\$se\_OR)

# Histogram of boots\$se\_OR

