

# Econ 293 Final Project Analysis Report

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```
# Set up data and global variables
source("helpers.R")
source("load-data.R")
data_01 <- split_control_treated(0, 1, df, covariate_names)
data_02 <- split_control_treated(0, 2, df, covariate_names)
data_03 <- split_control_treated(0, 3, df, covariate_names)
NUM TREES <- 1000

# Estimate HTEs using forest-based learners
if (!file.exists("results/results_forests.rds")) {
  set.seed(123)
  results_forests_01 <- compare_forests(data_01$X, data_01$W, data_01$Y, data_01$ID, NUM TREES)
  results_forests_01$compare <- "0 vs 1"

  results_forests_02 <- compare_forests(data_02$X, data_02$W, data_02$Y, data_02$ID, NUM TREES)
  results_forests_02$compare <- "0 vs 2"

  results_forests_03 <- compare_forests(data_03$X, data_03$W, data_03$Y, data_03$ID, NUM TREES)
  results_forests_03$compare <- "0 vs 3"

  results_forests <- Reduce(rbind, list(results_forests_01,
                                         results_forests_02,
                                         results_forests_03))
  saveRDS(results_forests, file = "results/results_forests.rds")
} else {
  results_forests <- readRDS(file = "results/results_forests.rds")
}

# Estimate HTEs using ridge-based learners
if (!file.exists("results/results_glmnets.rds")) {
  set.seed(1234)
  results_glmnets_01 <- compare_glmnets(data_01$X, data_01$W, data_01$Y, data_01$ID, 10, 0, NUM TREES)
  results_glmnets_01$compare <- "0 vs 1"

  results_glmnets_02 <- compare_glmnets(data_02$X, data_02$W, data_02$Y, data_02$ID, 10, 0, NUM TREES)
  results_glmnets_02$compare <- "0 vs 2"

  results_glmnets_03 <- compare_glmnets(data_03$X, data_03$W, data_03$Y, data_03$ID, 10, 0, NUM TREES)
  results_glmnets_03$compare <- "0 vs 3"

  results_glmnets <- Reduce(rbind, list(results_glmnets_01,
                                         results_glmnets_02,
                                         results_glmnets_03))
  saveRDS(results_glmnets, file = "results/results_glmnets.rds")
} else {
  results_glmnets <- readRDS(file = "results/results_glmnets.rds")
```

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}

# Combine all the results
hte_results <- rbind(results_forests, results_glmnets)

# Summarize the learner results
hte_summary_01 <- summarize_hte_results(hte_results, "0 vs 1")
hte_summary_02 <- summarize_hte_results(hte_results, "0 vs 2")
hte_summary_03 <- summarize_hte_results(hte_results, "0 vs 3")

# Print out the results
cat("0 vs 1", "\n")

## 0 vs 1

hte_summary_01$MRL <- scientific(hte_summary_01$MRL)
hte_summary_01$SDRL <- scientific(hte_summary_01$SDRL)
print(hte_summary_01)

## # A tibble: 7 x 3
##   learner      MRL      SDRL
##   <fct>       <chr>    <chr>
## 1 S-forest    1.32e-04  1.64e-03
## 2 T-forest    1.36e-05  7.03e-04
## 3 X-forest    8.47e-06  1.99e-03
## 4 Causal forest 3.47e-05  1.80e-03
## 5 S-glmnet    1.56e-06  2.32e-04
## 6 T-glmnet    9.63e-07  2.19e-04
## 7 X-glmnet    -8.64e-07 1.37e-04

cat("\n\n\n")

cat("0 vs 2", "\n")

## 0 vs 2

hte_summary_02$MRL <- scientific(hte_summary_02$MRL)
hte_summary_02$SDRL <- scientific(hte_summary_02$SDRL)
print(hte_summary_02)

## # A tibble: 7 x 3
##   learner      MRL      SDRL
##   <fct>       <chr>    <chr>
## 1 S-forest    1.03e-04  1.35e-03
## 2 T-forest    1.32e-05  7.31e-04
## 3 X-forest    4.97e-06  1.97e-03
## 4 Causal forest 3.04e-05  1.87e-03
## 5 S-glmnet    2.47e-06  4.04e-04
## 6 T-glmnet    8.62e-07  2.64e-04
## 7 X-glmnet    4.63e-07  1.44e-04

cat("\n\n\n")

cat("0 vs 3", "\n")

## 0 vs 3

```

```

hte_summary_03$MRL <- scientific(hte_summary_03$MRL)
hte_summary_03$SDRL <- scientific(hte_summary_03$SDRL)
print(hte_summary_03)

## # A tibble: 7 x 3
##   learner      MRL      SDRL
##   <fct>       <chr>    <chr>
## 1 S-forest    9.28e-05 1.20e-03
## 2 T-forest    3.62e-06 7.51e-04
## 3 X-forest    -1.67e-05 2.20e-03
## 4 Causal forest 1.24e-05 2.08e-03
## 5 S-glmnet    2.13e-06 4.42e-04
## 6 T-glmnet    2.59e-07 2.79e-04
## 7 X-glmnet    4.51e-06 3.21e-04

cat("\n\n\n")

# Find the best learners
hte_best_01 <- hte_results %>%
  filter(compare == "0 vs 1") %>%
  filter(learner == hte_summary_01$learner[hte_summary_01$MRL == min(hte_summary_01$MRL)])
hte_best_learner_01 <- unique(hte_best_01$learner)
print(paste0("Best learner for 0 vs 1: ", hte_best_learner_01))

## [1] "Best learner for 0 vs 1: X-glmnet"

hte_best_02 <- hte_results %>%
  filter(compare == "0 vs 2") %>%
  filter(learner == hte_summary_02$learner[hte_summary_02$MRL == min(hte_summary_02$MRL)])
hte_best_learner_02 <- unique(hte_best_02$learner)
print(paste0("Best learner for 0 vs 2: ", hte_best_learner_02))

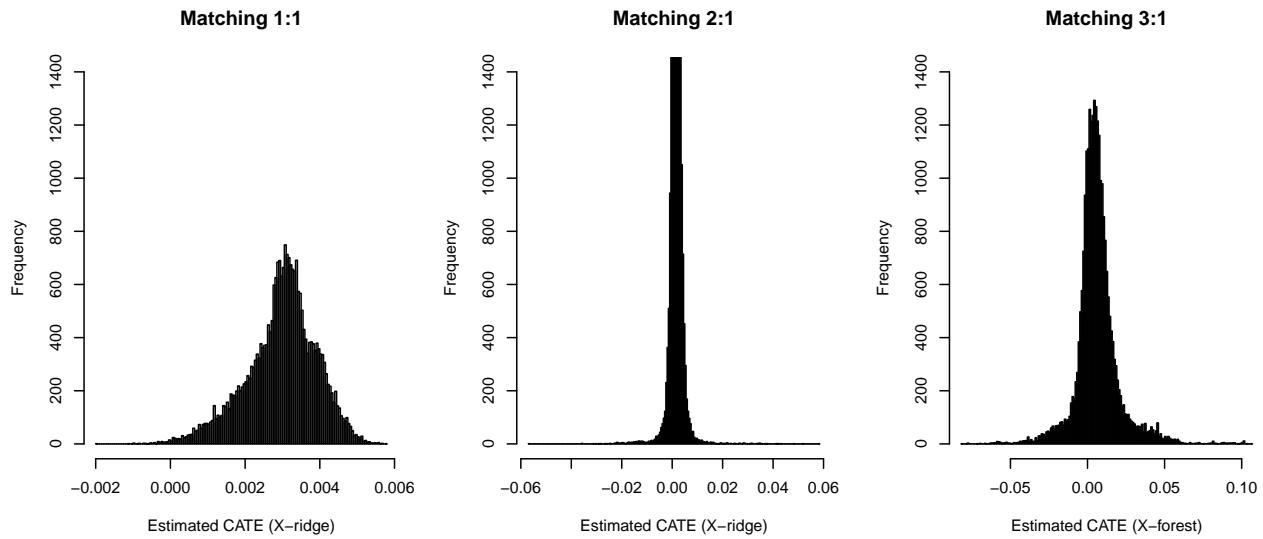
## [1] "Best learner for 0 vs 2: S-forest"

hte_best_03 <- hte_results %>%
  filter(compare == "0 vs 3") %>%
  filter(learner == hte_summary_03$learner[hte_summary_03$MRL == min(hte_summary_03$MRL)])
hte_best_leanrer_03 <- unique(hte_best_03$learner)
print(paste0("Best learner for 0 vs 3: ", hte_best_leanrer_03))

## [1] "Best learner for 0 vs 3: X-forest"

par(mfrow = c(1, 3))
hist(hte_best_01$estimate, breaks = 250, ylim = c(0, 1400), xlab = "Estimated CATE (X-ridge)", main = "X-ridge")
hist(hte_best_02$estimate, breaks = 250, ylim = c(0, 1400), xlab = "Estimated CATE (X-ridge)", main = "X-ridge")
hist(hte_best_03$estimate, breaks = 250, ylim = c(0, 1400), xlab = "Estimated CATE (X-forest)", main = "X-forest")

```



```
# Hypothesis test for overall treatment heterogeneity
if (!file.exists("results/test_hetero_summary_01.rds")) {
  test_hetero_01 <- test_hetero(as.matrix(hte_best_01[, covariate_names]), hte_best_01$W,
                                hte_best_01$Y, hte_best_01$estimate, NUM TREES)
  test_hetero_summary_01 <- summarize_test_hetero(test_hetero_01)
  saveRDS(test_hetero_summary_01, file = "results/test_hetero_summary_01.rds")
} else {
  test_hetero_summary_01 <- readRDS("results/test_hetero_summary_01.rds")
}

if (!file.exists("results/test_hetero_summary_02.rds")) {
  test_hetero_02 <- test_hetero(as.matrix(hte_best_02[, covariate_names]), hte_best_02$W,
                                hte_best_02$Y, hte_best_02$estimate, NUM TREES)
  test_hetero_summary_02 <- summarize_test_hetero(test_hetero_02)
  saveRDS(test_hetero_summary_02, file = "results/test_hetero_summary_02.rds")
} else {
  test_hetero_summary_02 <- readRDS("results/test_hetero_summary_02.rds")
}

if (!file.exists("results/test_hetero_summary_03.rds")) {
  test_hetero_03 <- test_hetero(as.matrix(hte_best_03[, covariate_names]), hte_best_03$W,
                                hte_best_03$Y, hte_best_03$estimate, NUM TREES)
  test_hetero_summary_03 <- summarize_test_hetero(test_hetero_03)
  saveRDS(test_hetero_summary_03, file = "results/test_hetero_summary_03.rds")
} else {
  test_hetero_summary_03 <- readRDS("results/test_hetero_summary_03.rds")
}

cat("Test for overall heterogeneity", "\n")

## Test for overall heterogeneity
cat("0 vs 1", "\n")

## 0 vs 1
```

```

test_hetero_summary_01$onesided_pval <- min(c(1, test_hetero_summary_01$onesided_pval * 3))
print(test_hetero_summary_01)

## $beta
## [1] 2.602721
##
## $beta_se
## [1] 1.822936
##
## $onesided_pval
## [1] 0.2300578
cat("\n\n")

cat("0 vs 2", "\n")

## 0 vs 2

test_hetero_summary_02$onesided_pval <- min(c(1, test_hetero_summary_02$onesided_pval * 3))
print(test_hetero_summary_02)

## $beta
## [1] -0.4299797
##
## $beta_se
## [1] 1.818228
##
## $onesided_pval
## [1] 1
cat("\n\n")

cat("0 vs 3", "\n")

## 0 vs 3

test_hetero_summary_03$onesided_pval <- min(c(1, test_hetero_summary_03$onesided_pval * 3))
print(test_hetero_summary_03)

## $beta
## [1] 0.6757758
##
## $beta_se
## [1] 0.1246052
##
## $onesided_pval
## [1] 8.852597e-08
cat("\n\n")

# Fit the models and get results
donate_cost_ratio <- c(10, 100, 1000, 10000)
costs <- 1 / donate_cost_ratio
if (!file.exists("results/policy_results_03.rds")) {
  set.seed(123)
  policy_results_03 <- list()
  for (i in 1:length(costs)) {
    policy_results_03[[i]] <- opt_policy_forest(data_03$X, data_03$W, data_03$Y, hte_best_03$estimate,
  }
}

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cost = costs[i], k = 10, num_trees = NUM_TREES)
}
saveRDS(policy_results_03, file = "results/policy_results_03.rds")
} else {
  policy_results_03 <- readRDS("results/policy_results_03.rds")
}

# Result summaries
policy_summary_dfs <- list()
for (i in 1:length(costs)) {
  policy_summary_dfs[[i]] <- summarize_opt_policy(policy_results_03[[i]]$Ahats_test)
  print(paste0("Donation:cost ratio = ", donate_cost_ratio[i]))
  print(policy_summary_dfs[[i]])
}

## [1] "Donation:cost ratio = 10"
##   Ahat_policy_mean Ahat_policy_sd Ahat_plugin_mean Ahat_plugin_sd
## 1      0.09381904    0.002706898     0.09382653    0.002713448
##   Ahat_everyone_mean Ahat_everyone_sd
## 1      -0.09381904    0.002706898
## [1] "Donation:cost ratio = 100"
##   Ahat_policy_mean Ahat_policy_sd Ahat_plugin_mean Ahat_plugin_sd
## 1      0.00401501    0.004897653     0.0104208    0.00399413
##   Ahat_everyone_mean Ahat_everyone_sd
## 1      -0.003815493   0.005984501
## [1] "Donation:cost ratio = 1000"
##   Ahat_policy_mean Ahat_policy_sd Ahat_plugin_mean Ahat_plugin_sd
## 1      0.004703359   0.004493456     0.009873607   0.005088723
##   Ahat_everyone_mean Ahat_everyone_sd
## 1      0.005160319   0.003941542
## [1] "Donation:cost ratio = 10000"
##   Ahat_policy_mean Ahat_policy_sd Ahat_plugin_mean Ahat_plugin_sd
## 1      0.005841853   0.002113163     0.009968833   0.00231166
##   Ahat_everyone_mean Ahat_everyone_sd
## 1      0.006024226   0.002224656

df1 <- data_01$df %>% select(-ID)
df2 <- data_02$df %>% select(-ID)
df3 <- data_03$df %>% select(-ID)

if (!file.exists("results/ate_results_table.rds")) {
  ate_results <- calculate_ATE(df1, df2, df3, 1)
  saveRDS(ate_results, file = "results/ate_results_table.rds")
} else {
  ate_results <- readRDS("results/ate_results_table.rds")
}

if (!file.exists("results/ate_results_reduced_table.rds")) {
  ate_results_reduced <- calculate_ATE(df1, df2, df3, 0.5)
  saveRDS(ate_results_reduced, file = "results/ate_results_reduced_table.rds")
} else {
  ate_results_reduced <- readRDS("results/ate_results_reduced_table.rds")
}
print(ate_results)

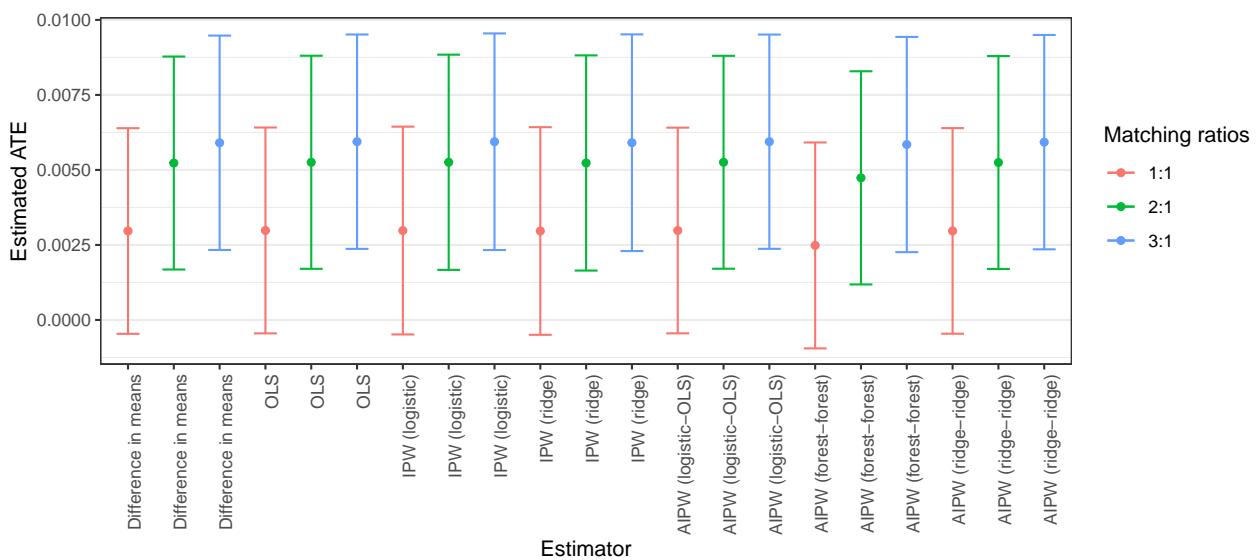
```

```

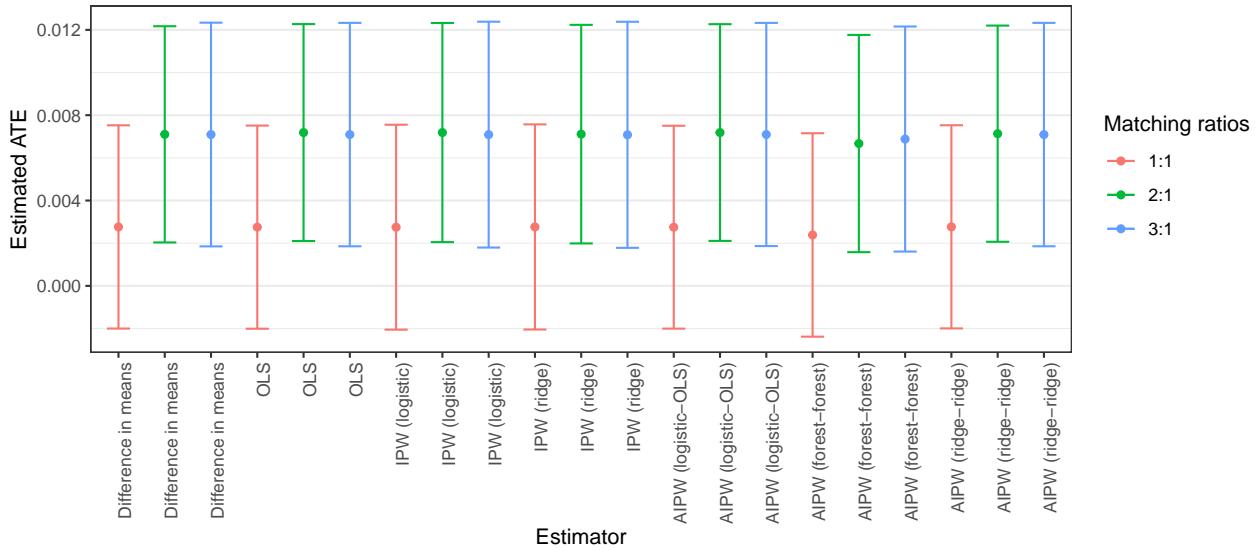
## # A tibble: 21 x 5
##   Estimator          ATE  lower_ci upper_ci treatment_lvl
##   <chr>            <dbl>    <dbl>    <dbl>    <chr>
## 1 RCT_gold_standard_1 0.00296 -0.000462  0.00639  1
## 2 RCT_gold_standard_2 0.00523  0.00168   0.00878  2
## 3 RCT_gold_standard_3 0.00590  0.00233   0.00948  3
## 4 linear_regression_1 0.00298 -0.000447  0.00641  1
## 5 linear_regression_2 0.00526  0.00170   0.00881  2
## 6 linear_regression_3 0.00594  0.00237   0.00951  3
## 7 IPW_logistic_1      0.00298 -0.000482  0.00644  1
## 8 IPW_logistic_2      0.00526  0.00167   0.00884  2
## 9 IPW_logistic_3      0.00594  0.00233   0.00955  3
## 10 IPW_glmnet_1       0.00297 -0.000495  0.00643  1
## # ... with 11 more rows
print(ate_results_reduced)

## # A tibble: 21 x 5
##   Estimator          ATE  lower_ci upper_ci treatment_lvl
##   <chr>            <dbl>    <dbl>    <dbl>    <chr>
## 1 RCT_gold_standard_1 0.00276 -0.00200  0.00753  1
## 2 RCT_gold_standard_2 0.00710  0.00204  0.0122   2
## 3 RCT_gold_standard_3 0.00709  0.00185  0.0123   3
## 4 linear_regression_1 0.00275 -0.00201  0.00751  1
## 5 linear_regression_2 0.00719  0.00210  0.0123   2
## 6 linear_regression_3 0.00709  0.00185  0.0123   3
## 7 IPW_logistic_1      0.00275 -0.00205  0.00755  1
## 8 IPW_logistic_2      0.00719  0.00205  0.0123   2
## 9 IPW_logistic_3      0.00709  0.00180  0.0124   3
## 10 IPW_glmnet_1       0.00276 -0.00205  0.00757  1
## # ... with 11 more rows
ate_results$Estimator <- factor(ate_results$Estimator, levels = ate_results$Estimator)
ate_results_reduced$Estimator <- factor(ate_results_reduced$Estimator, levels = ate_results_reduced$Estimator)
plot_ATE_estimates(ate_results)

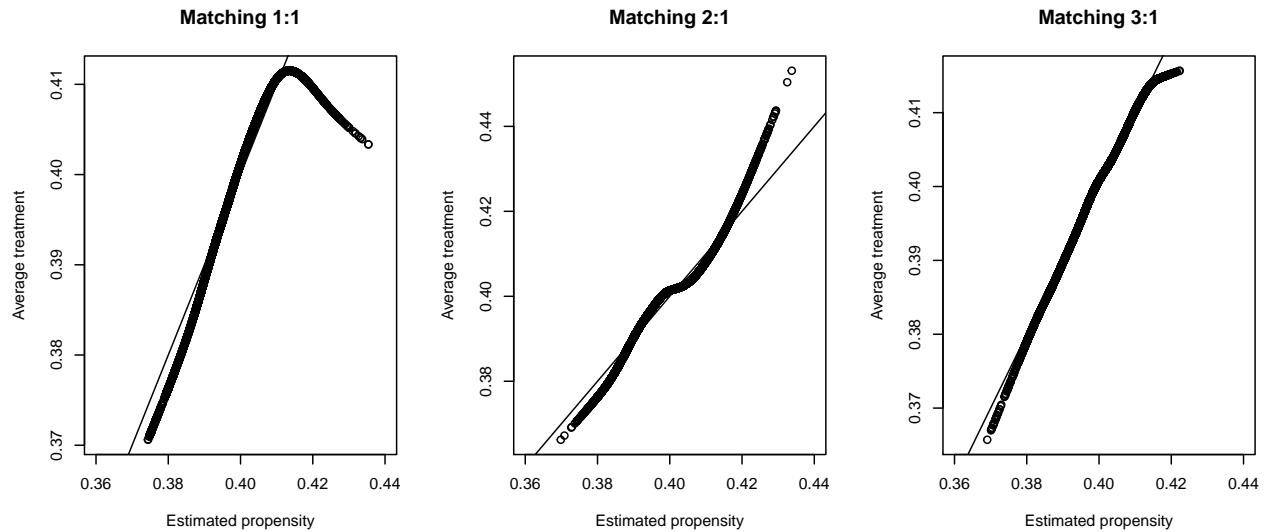
```



```
plot_ATE_estimates(ate_results_reduced)
```



```
p_logistic1 = p_logistic(df1)
p_logistic2 = p_logistic(df2)
p_logistic3 = p_logistic(df3)
par(mfrow = c(1, 3))
plot_calibration(p_logistic1, df1$W, ylab = "Average treatment", xlab = "Estimated propensity",
                 xlim = c(0.36, 0.44), main = "Matching 1:1")
plot_calibration(p_logistic2, df2$W, ylab = "Average treatment", xlab = "Estimated propensity",
                 xlim = c(0.36, 0.44), main = "Matching 2:1")
plot_calibration(p_logistic3, df3$W, ylab = "Average treatment", xlab = "Estimated propensity",
                 xlim = c(0.36, 0.44), main = "Matching 3:1")
```



```
par(mfrow=c(3,3))
plot_yx(p_logistic(data_01$df), data_01$df$pblack, "Estimated propensity", "Proportion black", main = "P")
plot_yx(p_logistic(data_01$df), data_01$df$median_hhincome, "Estimated propensity", "Median household income", main = "P")
plot_yx(p_logistic(data_01$df), as.factor(data_01$df$red0), "Estimated propensity", "Living in red state", names = c("No", "Yes"))

plot_yx(p_logistic(data_02$df), data_02$df$pblack, "Estimated propensity", "Proportion black", main = "P")
plot_yx(p_logistic(data_02$df), data_02$df$median_hhincome, "Estimated propensity", "Median household income", main = "P")
```

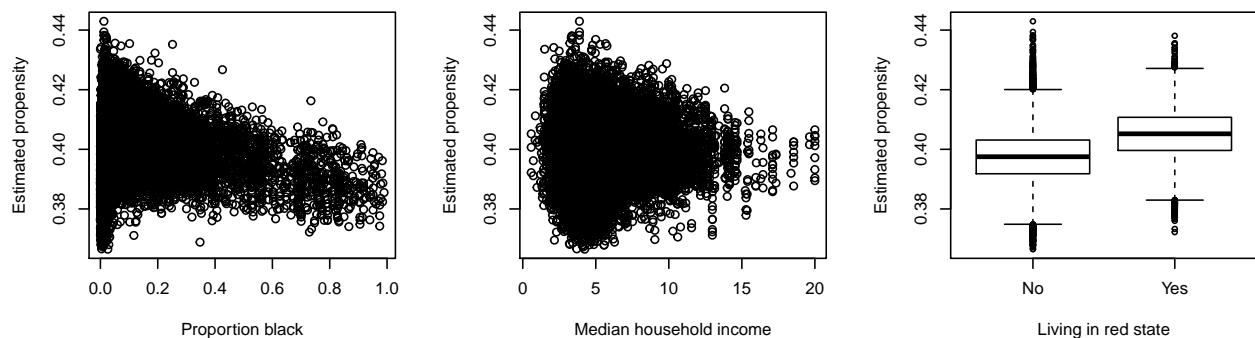
```

plot_yx(p_logistic(data_02$df), as.factor(data_02$df$red0), "Estimated propensity", "Living in red state"
        names = c("No", "Yes"))

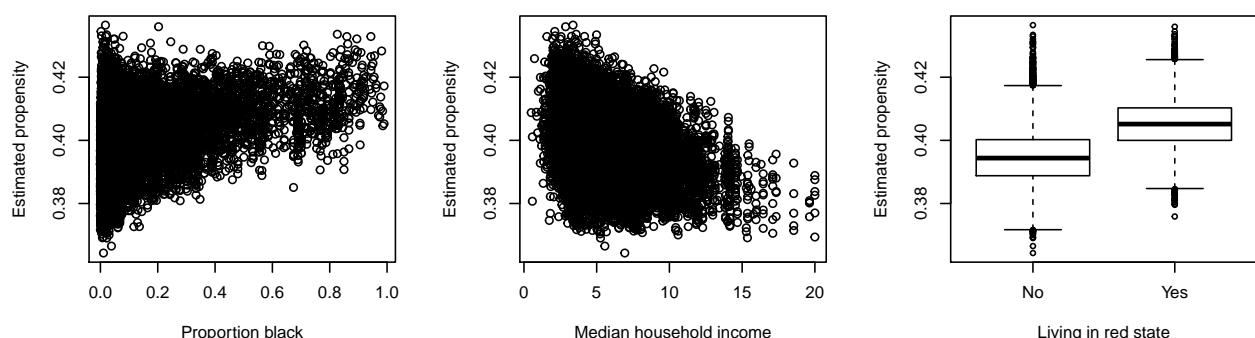
plot_yx(p_logistic(data_03$df), data_03$df$pblack, "Estimated propensity", "Proportion black", main = "Matching 1:1")
plot_yx(p_logistic(data_03$df), data_03$df$median_hhincome, "Estimated propensity", "Median household income", main = "Matching 1:1")
plot_yx(p_logistic(data_03$df), as.factor(data_03$df$red0), "Estimated propensity", "Living in red state", main = "Matching 1:1")
        names = c("No", "Yes"))

```

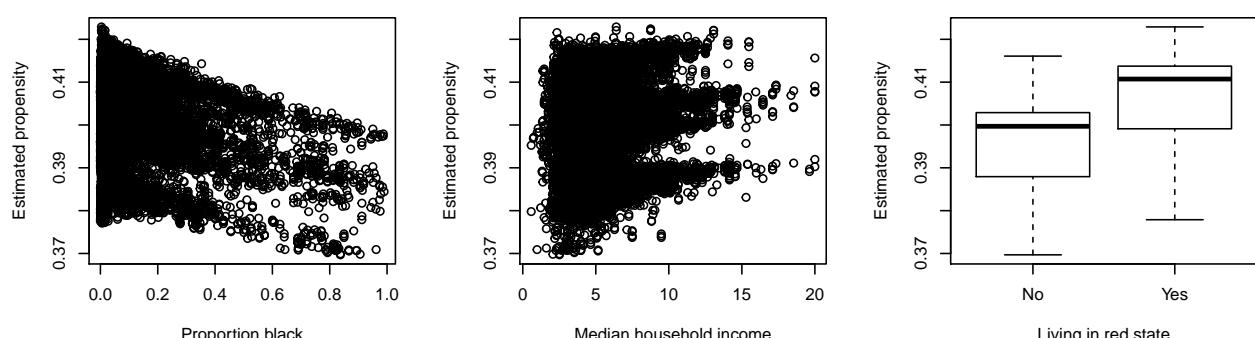
**Matching 1:1**



**Matching 2:1**



**Matching 3:1**



```

par(mfrow=c(3,3))
plot_yx(hte_best_01$estimate, hte_best_01$pblack, "Estimated CATE (X-ridge)", "Proportion black", main = "Matching 1:1")
plot_yx(hte_best_01$estimate, hte_best_01$median_hhincome, "Estimated CATE (X-ridge)", "Median household income", main = "Matching 1:1")
plot_yx(hte_best_01$estimate, as.factor(hte_best_01$red0), "Estimated CATE (X-ridge)", "Living in red state", main = "Matching 1:1")
        names = c("No", "Yes"))

```

```

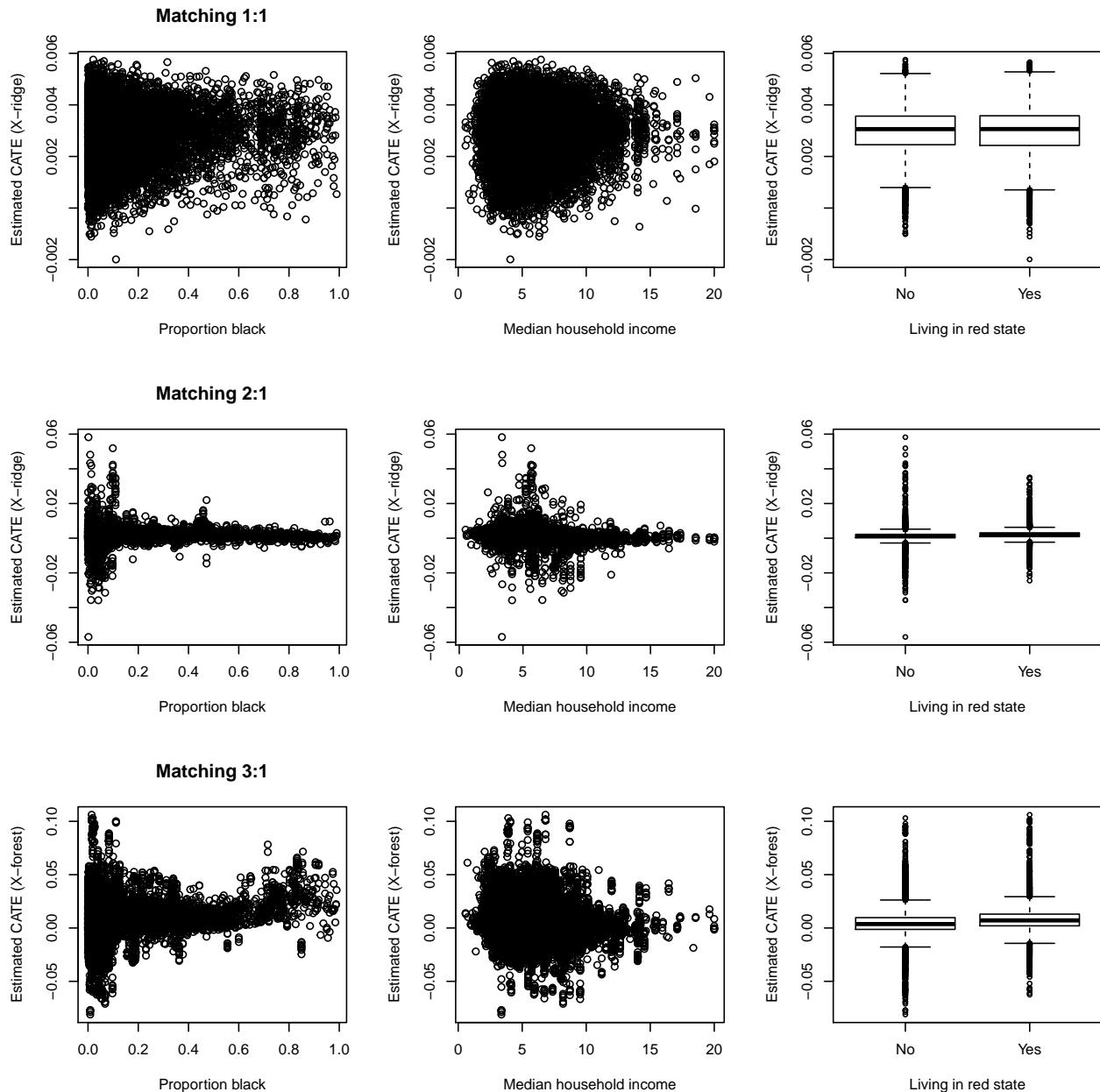
plot_yx(hte_best_02$estimate, hte_best_02$pblack, "Estimated CATE (X-ridge)", "Proportion black", main = "Matching 1:1")
plot_yx(hte_best_02$estimate, hte_best_02$median_hhincome, "Estimated CATE (X-ridge)", "Median household income", main = "Matching 1:1")
plot_yx(hte_best_02$estimate, as.factor(hte_best_02$red0), "Estimated CATE (X-ridge)", "Living in red state", main = "Matching 1:1")
names = c("No", "Yes"))

```

```

plot_yx(hte_best_03$estimate, hte_best_03$pblack, "Estimated CATE (X-forest)", "Proportion black", main = "Matching 2:1")
plot_yx(hte_best_03$estimate, hte_best_03$median_hhincome, "Estimated CATE (X-forest)", "Median household income", main = "Matching 2:1")
plot_yx(hte_best_03$estimate, as.factor(hte_best_03$red0), "Estimated CATE (X-forest)", "Living in red state", main = "Matching 2:1")
names = c("No", "Yes"))

```



```

# Compute the doubly robust scores
if (!file.exists("results/hte_best_01.rds")) {
  drs_best_01 <- estimate_binary_drs(X_train = as.matrix(hte_best_01[, covariate_names]),
                                         W_train = hte_best_01$W,
                                         W_val = hte_best_01$W_val,
                                         W_test = hte_best_01$W_test)
}

```

```

        Y_train = hte_best_01$Y,
        cate_train = hte_best_01$estimate,
        X_new = NULL, W_new = NULL, Y_new = NULL, cate_new = NULL,
        num_trees = NUM TREES, compute_new = FALSE)

hte_best_01$drs <- drs_best_01$train
saveRDS(hte_best_01, file = "results/hte_best_01.rds")
} else {
  hte_best_01 <- readRDS("results/hte_best_01.rds")
}

if (!file.exists("results/hte_best_02.rds")) {
  drs_best_02 <- estimate_binary_drs(X_train = as.matrix(hte_best_02[ , covariate_names]),
                                      W_train = hte_best_02$W,
                                      Y_train = hte_best_02$Y,
                                      cate_train = hte_best_02$estimate,
                                      X_new = NULL, W_new = NULL, Y_new = NULL, cate_new = NULL,
                                      num_trees = NUM TREES, compute_new = FALSE)

  hte_best_02$drs <- drs_best_02$train
  saveRDS(hte_best_02, file = "results/hte_best_02.rds")
} else {
  hte_best_02 <- readRDS("results/hte_best_02.rds")
}

if (!file.exists("results/hte_best_03.rds")) {
  drs_best_03 <- estimate_binary_drs(X_train = as.matrix(hte_best_03[ , covariate_names]),
                                      W_train = hte_best_03$W,
                                      Y_train = hte_best_03$Y,
                                      cate_train = hte_best_03$estimate,
                                      X_new = NULL, W_new = NULL, Y_new = NULL, cate_new = NULL,
                                      num_trees = NUM TREES, compute_new = FALSE)

  hte_best_03$drs <- drs_best_03$train
  saveRDS(hte_best_03, file = "results/hte_best_03.rds")
} else {
  hte_best_03 <- readRDS("results/hte_best_03.rds")
}

# Formal hypothesis tests
test_covs <- c("pblack", "median_hhincome", "red0")
test_cov_types <- c("continuous", "continuous", "binary")
df_lst <- list("0 vs 1" = hte_best_01, "0 vs 2" = hte_best_02, "0 vs 3" = hte_best_03)
for (comparison in names(df_lst)) {
  for (i in 1:length(test_covs)) {
    cat(paste0("Comparison: ", comparison), "\n")
    cat(paste0("Covariate: ", test_covs[i]), "\n")
    t_test <- test_covariate_hetero(df = df_lst[[comparison]],
                                      x_name = test_covs[i],
                                      x_type = test_cov_types[i])
    print(t_test)
    cat("Corrected p-value: ", min(c(t_test$p.value*length(test_covs)*length(df_lst), 1)), "\n\n")
  }
  cat("\n\n\n")
}

## Comparison: 0 vs 1

```

```

## Covariate: pblack
##
## Welch Two Sample t-test
##
## data: drs[!high_idx] and drs[high_idx]
## t = 0.42863, df = 26059, p-value = 0.6682
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.003515538 0.005483472
## sample estimates:
## mean of x mean of y
## 0.003288620 0.002304653
##
## Corrected p-value: 1
##
## Comparison: 0 vs 1
## Covariate: median_hhincome
##
## Welch Two Sample t-test
##
## data: drs[!high_idx] and drs[high_idx]
## t = 0.21332, df = 26003, p-value = 0.8311
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.004009763 0.004989139
## sample estimates:
## mean of x mean of y
## 0.003043472 0.002553785
##
## Corrected p-value: 1
##
## Comparison: 0 vs 1
## Covariate: red0
##
## Welch Two Sample t-test
##
## data: drs[!pos_idx] and drs[pos_idx]
## t = -1.9775, df = 24229, p-value = 0.048
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -8.961753e-03 -3.960099e-05
## sample estimates:
## mean of x mean of y
## 0.001014136 0.005514814
##
## Corrected p-value: 0.4320074
##
##
##
##
## Comparison: 0 vs 2
## Covariate: pblack
##
## Welch Two Sample t-test

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##
## data: drs[!high_idx] and drs[high_idx]
## t = -0.0069828, df = 25988, p-value = 0.9944
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.004520740 0.004488643
## sample estimates:
##   mean of x   mean of y
## 0.005214524 0.005230572
##
## Corrected p-value: 1
##
## Comparison: 0 vs 2
## Covariate: median_hhincome
##
## Welch Two Sample t-test
##
## data: drs[!high_idx] and drs[high_idx]
## t = 0.60634, df = 25920, p-value = 0.5443
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.003111179 0.005898235
## sample estimates:
##   mean of x   mean of y
## 0.005919177 0.004525650
##
## Corrected p-value: 1
##
## Comparison: 0 vs 2
## Covariate: red0
##
## Welch Two Sample t-test
##
## data: drs[!pos_idx] and drs[pos_idx]
## t = -1.945, df = 23927, p-value = 0.05179
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.0089389086 0.0000344735
## sample estimates:
##   mean of x   mean of y
## 0.003451801 0.007904018
##
## Corrected p-value: 0.4660769
##
##
##
##
## Comparison: 0 vs 3
## Covariate: pblack
##
## Welch Two Sample t-test
##
## data: drs[!high_idx] and drs[high_idx]
## t = 0.19059, df = 26075, p-value = 0.8489

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## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.004073128  0.004950545
## sample estimates:
##   mean of x   mean of y
## 0.006350420 0.005911711
##
## Corrected p-value:  1
##
## Comparison: 0 vs 3
## Covariate: median_hhincome
##
## Welch Two Sample t-test
##
## data: drs[!high_idx] and drs[high_idx]
## t = 1.7588, df = 25978, p-value = 0.07862
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.0004631758  0.0085599777
## sample estimates:
##   mean of x   mean of y
## 0.008155300 0.004106899
##
## Corrected p-value:  0.7075657
##
## Comparison: 0 vs 3
## Covariate: red0
##
## Welch Two Sample t-test
##
## data: drs[!pos_idx] and drs[pos_idx]
## t = -2.9333, df = 23903, p-value = 0.003357
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.011237163 -0.002235001
## sample estimates:
##   mean of x   mean of y
## 0.003450408 0.010186490
##
## Corrected p-value:  0.03021009
# test_CATE(hte_best_01, "estimate", "pblack", "continuous")
# test_CATE(hte_best_01, "estimate", "median_hhincome", "continuous")
# test_CATE(hte_best_01, "estimate", "red0", "factor")
#
# test_CATE(hte_best_02, "estimate", "pblack", "continuous")
# test_CATE(hte_best_02, "estimate", "median_hhincome", "continuous")
# test_CATE(hte_best_02, "estimate", "red0", "factor")
#
# test_CATE(hte_best_03, "estimate", "pblack", "continuous")
# test_CATE(hte_best_03, "estimate", "median_hhincome", "continuous")
# test_CATE(hte_best_03, "estimate", "red0", "factor")

```