HW5

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2023-11-20

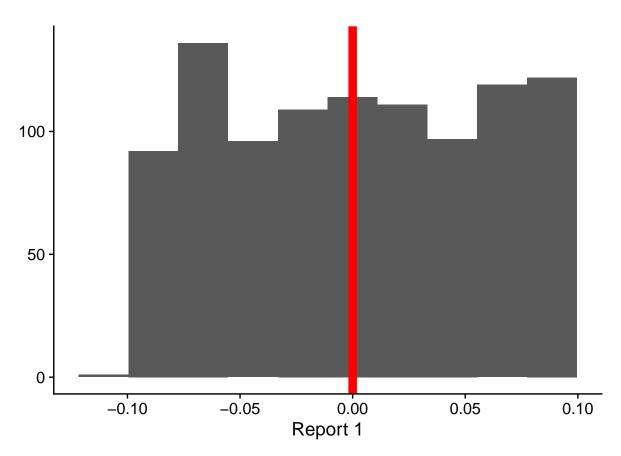
```
here::i_am("HW5/HW5.Rmd")
## here() starts at /Users/erikandersen/Documents/Classes/ECON 587
# Load packages
pacman::p_load(tidyverse, magrittr, rdd, ivreg)
# Load data
oz_df = haven::read_dta(here::here('HW5', "data", "Ozier_JHR_Econ587.dta"))
Question 1
# Generate indicator for treatment status. Treated if test > 0
oz_df = oz_df |> mutate(treatment = if_else(test > 0, 1, 0))
# Regress secondary on test score, treatment, and interaction
oz_df %>% lm(secondary ~ test*treatment,.) |> broom::tidy()
a)
## # A tibble: 4 x 5
           estimate std.error statistic p.value
   term
##
    <chr>
                    <dbl> <dbl> <dbl>
                                                 <dbl>
## 1 (Intercept)
                   0.330 0.0263 12.5 1.02e-34
                                       3.88 1.07e- 4
## 2 test
                            0.0688
                    0.267
## 3 treatment
                   0.165
                            0.0377
                                        4.38 1.24e- 5
## 4 test:treatment 0.0151 0.0967
                                         0.157 8.76e- 1
# Define restrictions
restrict = c(0.8, 0.4, 0.2, 0.1)
lapply(restrict, function(x){
  oz_df |> filter(abs(test) < x) %>%
   lm(secondary ~ test*treatment,.) |>
   broom::tidy()
})
c)
## [[1]]
## # A tibble: 4 x 5
##
    term
                estimate std.error statistic p.value
##
    <chr>>
                     <dbl>
                               <dbl>
                                         <dbl>
                                                 <dbl>
```

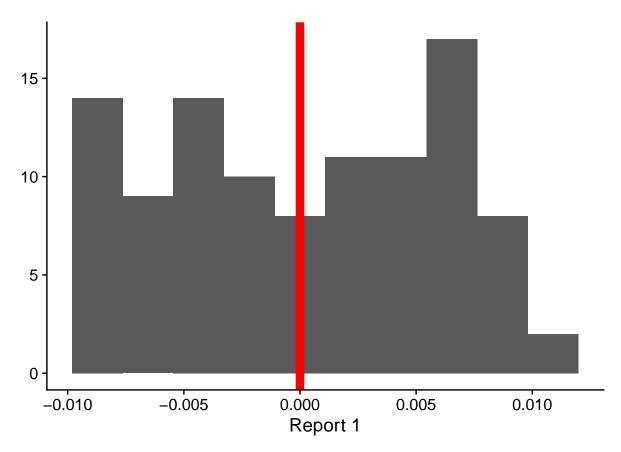
```
## 1 (Intercept)
                     0.330
                                0.0263
                                          12.5 1.02e-34
## 2 test
                      0.267
                                0.0688
                                           3.88 1.07e- 4
                                           4.38 1.24e- 5
## 3 treatment
                     0.165
                                0.0377
                                           0.157 8.76e- 1
## 4 test:treatment 0.0151
                                0.0967
## [[2]]
## # A tibble: 4 x 5
                    estimate std.error statistic p.value
##
    term
##
     <chr>>
                       <dbl>
                                 <dbl>
                                          <dbl>
                                                    <dbl>
                                0.0358
                                           8.56 3.28e-17
## 1 (Intercept)
                       0.306
## 2 test
                       0.117
                                0.168
                                           0.699 4.85e- 1
                                           3.57 3.71e- 4
## 3 treatment
                                0.0521
                       0.186
                                           0.783 4.34e- 1
## 4 test:treatment
                       0.186
                                0.237
##
## [[3]]
## # A tibble: 4 x 5
##
    term
                    estimate std.error statistic
                                                      p.value
##
     <chr>>
                       <dbl>
                                 <dbl>
                                           <dbl>
                                                        <dbl>
## 1 (Intercept)
                       0.283
                                0.0510
                                           5.55 0.0000000409
## 2 test
                      -0.125
                                0.462
                                          -0.271 0.786
## 3 treatment
                      0.237
                                0.0752
                                           3.15 0.00170
## 4 test:treatment
                       0.196
                                0.663
                                           0.295 0.768
##
## [[4]]
## # A tibble: 4 x 5
    term
                    estimate std.error statistic p.value
##
     <chr>>
                       <dbl>
                                <dbl>
                                          <dbl>
                                                    <dbl>
                       0.277
                                0.0724
                                           3.83 0.000153
## 1 (Intercept)
## 2 test
                      -0.281
                             1.26
                                          -0.223 0.824
## 3 treatment
                      0.224
                                0.114
                                          1.97 0.0498
## 4 test:treatment
                       0.506
                                1.99
                                           0.254 0.799
# Estimate rdd. Note default standard errors are heteroskedastic robust
RDestimate(secondary ~ test, oz_df, cutpoint = 0)
d)
##
## Call:
## RDestimate(formula = secondary ~ test, data = oz_df, cutpoint = 0)
## Coefficients:
##
       LATE
                Half-BW Double-BW
##
       0.181
                 0.227
                             0.171
# Plot rdd graph
oz_df |> group_by(a = cut(test, 26)) |>
  summarise(secondary=mean(secondary), test = mean(test)) |>
  ggplot(aes(test, secondary)) +
   geom_point() +
   geom_smooth(data = oz_df |> filter(test > 0), method = 'lm', col = 'black') +
   geom_smooth(data = oz_df |> filter(test < 0), method = 'lm', col = 'black') +</pre>
```

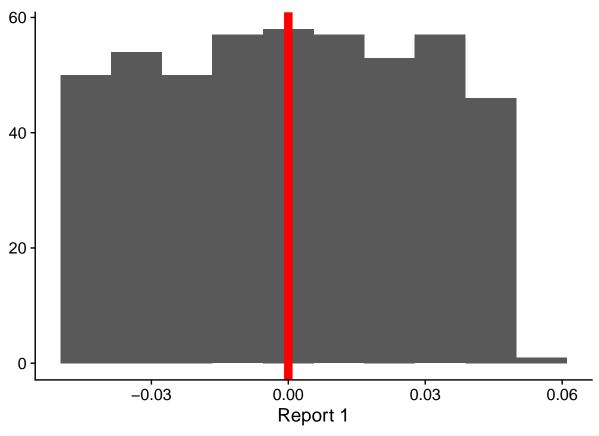
```
geom_vline(xintercept = 0, lty = 5) +
    xlab("Test Score") + ylab("Probability of Completing\n Secondary School") +
    cowplot::theme_cowplot()
e)
## `geom_smooth()` using formula = 'y ~ x'
## `geom_smooth()` using formula = 'y ~ x'
      8.0
Probability of Completing
Secondary School
      0.6
      0.4
      0.2
                                                                          0.5
                          -0.5
                                                  0.0
                                             Test Score
ggsave(here::here("HW5", "rdd_plot.pdf"))
## Saving 6.5 \times 4.5 in image
## `geom_smooth()` using formula = 'y ~ x'
## `geom_smooth()` using formula = 'y ~ x'
# Rerun esimation of rdd, but override default kernal and bandwidths
RDestimate(secondary ~ test, oz_df, cutpoint = 0, kernel = 'rectangular', bw = 0.8)
f)
##
## Call:
## RDestimate(formula = secondary ~ test, data = oz_df, cutpoint = 0,
       bw = 0.8, kernel = "rectangular")
##
##
## Coefficients:
##
        LATE
                 Half-BW Double-BW
       0.164
                   0.184
##
                               0.164
```

```
ivreg(rv ~ secondary + test + female + test:treatment |
        treatment + test + female + test:treatment, data = oz_df) |> broom::tidy()
\mathbf{g}
## # A tibble: 5 x 5
                  estimate std.error statistic p.value
    term
     <chr>
##
                      <dbl>
                                <dbl> <dbl>
                                                    <dbl>
## 1 (Intercept)
                      2.93
                               0.150
                                         19.5 1.64e-77
                      0.726 0.311
                                          2.33 1.98e- 2
## 2 secondary
                      0.612 0.159
                                          3.85 1.22e- 4
## 3 test
                                          -3.64 2.80e- 4
## 4 female
                     -0.176 0.0483
                                          -2.34 1.95e- 2
## 5 test:treatment
                    -0.311 0.133
# Run estimate again, but use controls. Note the regressors are in the opposite order than the homework
RDestimate(rv ~ test + secondary, oz df, cutpoint = 0)
h)
##
## RDestimate(formula = rv ~ test + secondary, data = oz_df, cutpoint = 0)
## Coefficients:
##
       LATE
               Half-BW Double-BW
##
        1.47
                   1.32
                             1.14
# Replicate q result. I don't include test interacted with treatment because it gives insane results li
RDestimate(rv ~ test + secondary | female , data = oz_df, cutpoint = 0,
           kernel = 'rectangular', bw = 0.6)
i)
##
## RDestimate(formula = rv ~ test + secondary | female, data = oz df,
       cutpoint = 0, bw = 0.6, kernel = "rectangular")
##
## Coefficients:
##
       LATE
              Half-BW Double-BW
##
       0.744
                 1.266
                            0.670
Question 2
# Load data
sim_df = haven::read_dta(here::here("HW5", "data", "RD_Manip_Econ587.dta"))
# Plot kernel densities for all four measures of wealth for bandwidths 0.1, 0.05, 0.01
lapply(c(0.1, 0.05, 0.01),
       function(x) {
         sim_df |> ggplot(aes(reportwealth1)) +
```

```
geom_density(bw = x) +
           geom_density(aes(reportwealth2), bw = x, col = 'red') +
           geom_density(aes(reportwealth3), bw = x, col = 'blue') +
           geom_density(aes(reportwealth4), bw = x, col = 'green') +
           xlab("Wealth Kernel Density") + ylab("") +
           ggtitle(paste("Wealth Kernel Density with Bandwidth =", x)) +
           labs(caption = "Black: Reported Wealth 1\n Red: Reported Wealth 2\n
                Blue: Reported Wealth 3\n Green: Reported Wealth 4") +
           cowplot::theme_cowplot()
         ggsave(here::here("HW5", paste(
           "Kernel_wealth_bandwidth_", x, ".pdf", sep = ""
         )))
       }
)
a)
## Saving 6.5 x 4.5 in image
## Saving 6.5 x 4.5 in image
## Saving 6.5 \times 4.5 in image
## [1] "/Users/erikandersen/Documents/Classes/ECON 587/HW5/Kernel_wealth_bandwidth_0.1.pdf"
##
## [1] "/Users/erikandersen/Documents/Classes/ECON 587/HW5/Kernel_wealth_bandwidth_0.05.pdf"
##
## [[3]]
## [1] "/Users/erikandersen/Documents/Classes/ECON 587/HW5/Kernel wealth bandwidth 0.01.pdf"
# Test smoothness around the cuttoff using ocular methods. Unfortunately to loop this over the four sce
bandwidths = c(0.1, 0.01, 0.05)
lapply(bandwidths, function(x){
  sim_df |> filter(abs(reportwealth1) < x) |>
    select(reportwealth1)|>
    ggplot(aes(reportwealth1)) +
        geom_histogram(bins = 10) +
        geom_vline(xintercept = 0, color = 'red', lwd = 3) +
        xlab("Report 1") + ylab("") + cowplot::theme cowplot()
})
b)
## [[1]]
```

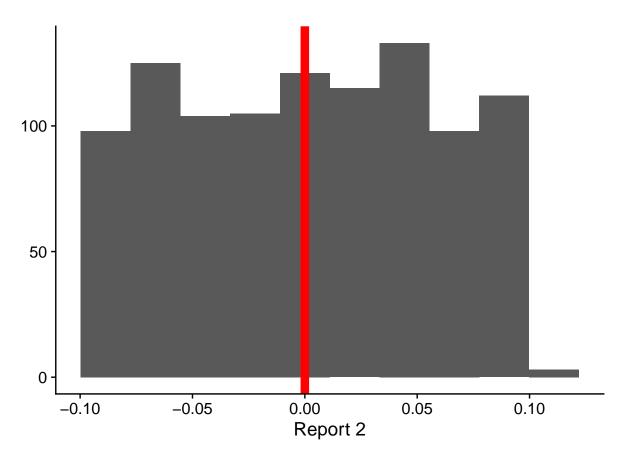


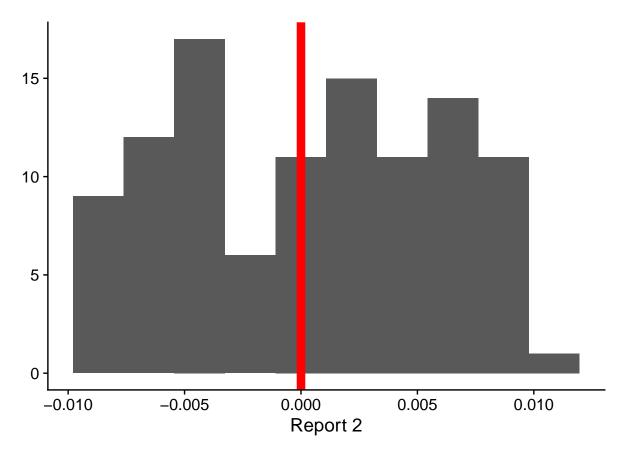


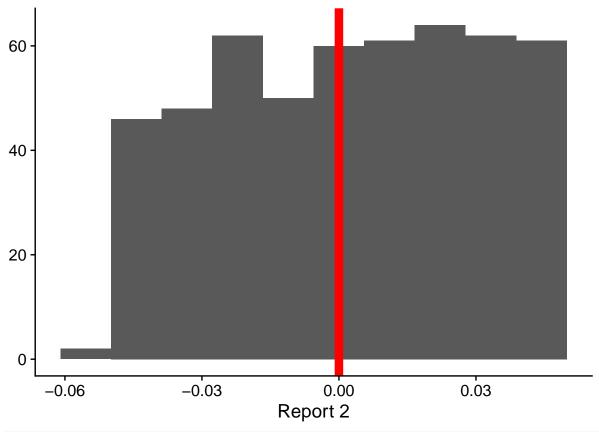


```
lapply(bandwidths, function(x){
    sim_df |> filter(abs(reportwealth2) < x) |>
    select(reportwealth2)|>
    ggplot(aes(reportwealth2)) +
        geom_histogram(bins = 10) +
        geom_vline(xintercept = 0, color = 'red', lwd = 3)+
        xlab("Report 2") + ylab("")+ cowplot::theme_cowplot()
})
```

[[1]]

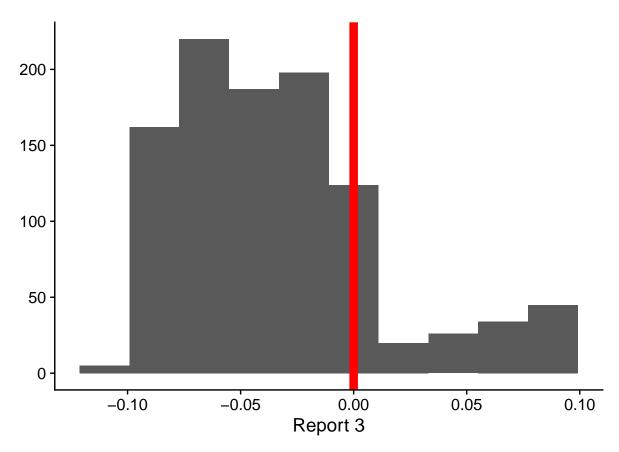


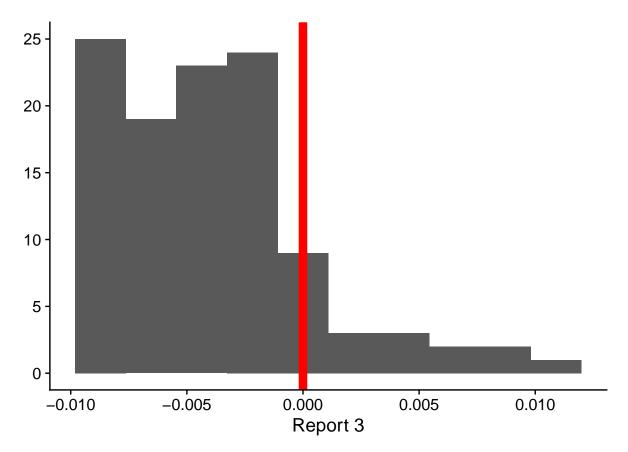


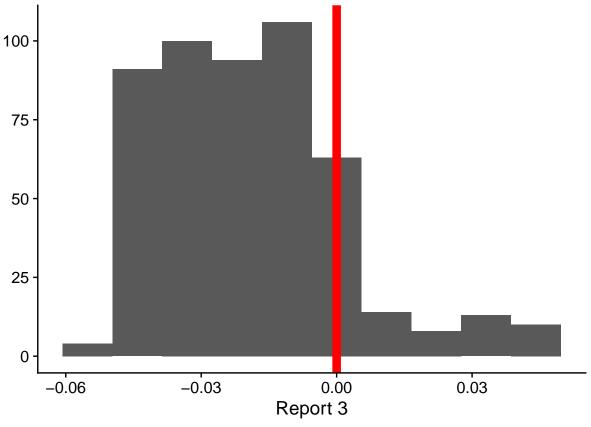


```
lapply(bandwidths, function(x){
    sim_df |> filter(abs(reportwealth3) < x) |>
    select(reportwealth3)|>
    ggplot(aes(reportwealth3)) +
        geom_histogram(bins = 10) +
        geom_vline(xintercept = 0, color = 'red', lwd = 3)+
        xlab("Report 3") + ylab("")+ cowplot::theme_cowplot()
})
```

[[1]]

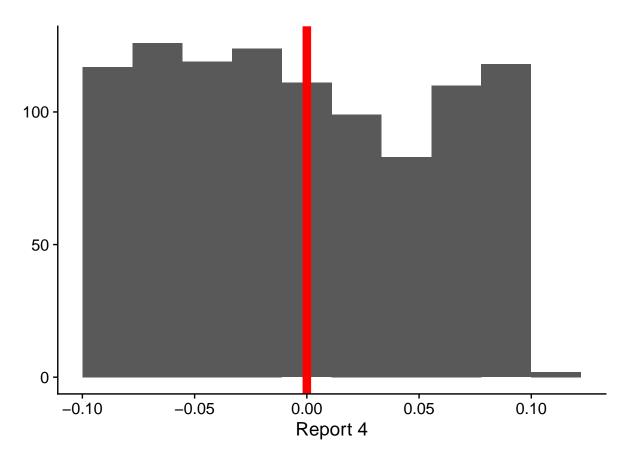


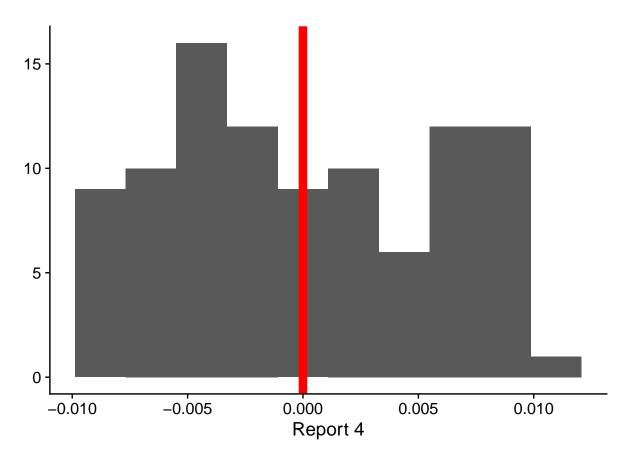


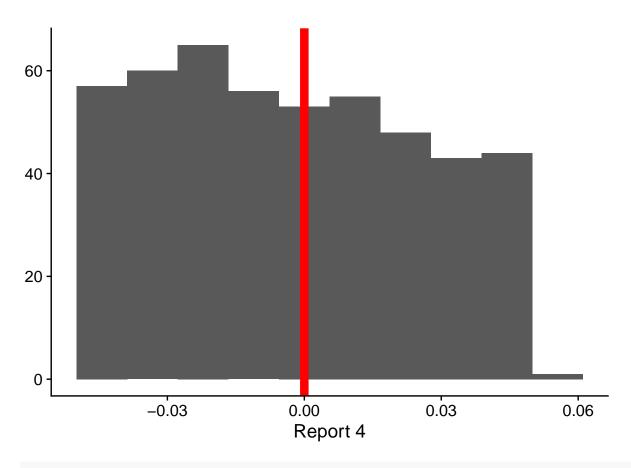


```
lapply(bandwidths, function(x){
  sim_df |> filter(abs(reportwealth4) < x) |>
    select(reportwealth4)|>
    ggplot(aes(reportwealth4)) +
        geom_histogram(bins = 10) +
        geom_vline(xintercept = 0, color = 'red', lwd = 3)+
        xlab("Report 4") + ylab("")+ cowplot::theme_cowplot()
})
```

[[1]]

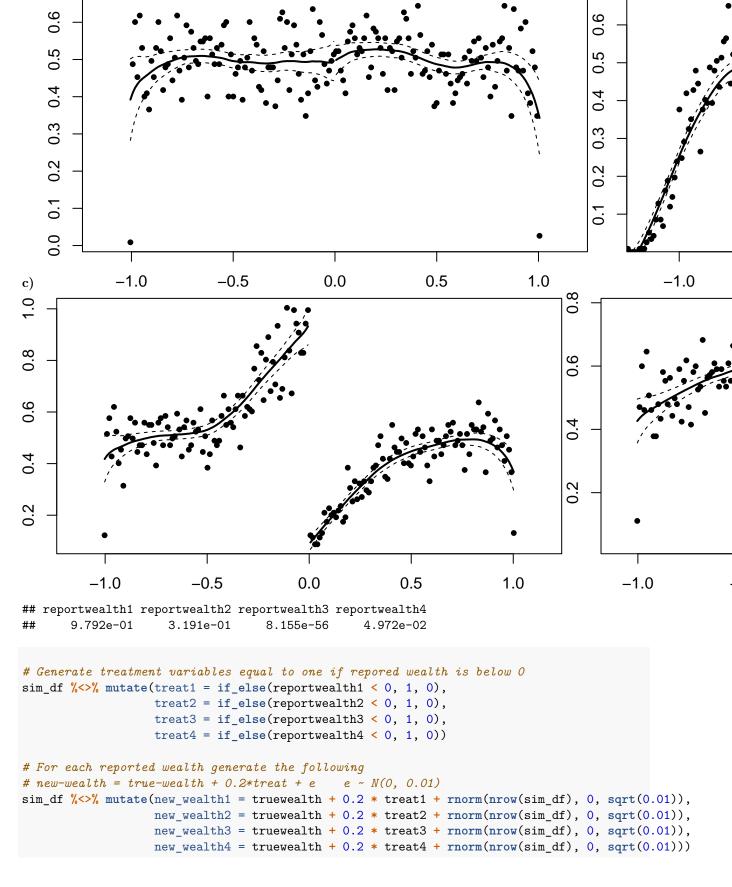






```
# McCrary kovac test
vars = paste0("reportwealth", 1:4)

# P-values are reported
sapply(vars, function(x){
    DCdensity(sim_df[[x]], cutpoint = 0)
})
```



```
# Summarize the new wealths
lapply(1:4, function(i){
  pdf(here::here("HW5", paste0("new-wealth", i, ".pdf")))
  plot(c(sim_df[paste0("reportwealth", i)]) |> unlist(),
       c(sim_df[paste0('new_wealth', i)]) |> unlist(),
       xlab = paste0("Report Wealth", i),
       ylab = paste0("New Wealth", i))
  dev.off()
})
d)
## [[1]]
## pdf
##
##
## [[2]]
## pdf
##
##
## [[3]]
## pdf
##
##
## [[4]]
## pdf
lapply(1:4, function(i){
  # Generate formula
  y = paste0("new_wealth", i)
  treat = paste0("treat", i)
  report = paste0("reportwealth", i)
  x = paste(treat, report, sep = "*")
  formula = paste(y, x, sep = "~")
  # Run regression
  lm(formula, sim_df) |> summary()
})
e)
## [[1]]
##
## Call:
## lm(formula = formula, data = sim_df)
## Residuals:
                1Q Median
                                3Q
## -0.3850 -0.0685 0.0008 0.0684 0.3467
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)
                       0.002187
                                  0.002767
                                              0.79
                                                        0.43
## treat1
                       0.197881
                                  0.003957
                                             50.01
                                                      <2e-16 ***
## reportwealth1
                       1.000628
                                  0.004853
                                            206.20
                                                      <2e-16 ***
## treat1:reportwealth1 0.000624
                                  0.006888
                                              0.09
                                                        0.93
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.0994 on 9996 degrees of freedom
## Multiple R-squared: 0.961, Adjusted R-squared: 0.961
## F-statistic: 8.16e+04 on 3 and 9996 DF, p-value: <2e-16
##
##
## [[2]]
##
## Call:
## lm(formula = formula, data = sim_df)
##
## Residuals:
##
               1Q Median
                               3Q
      Min
                                      Max
## -0.5107 -0.0953 0.0025 0.0964 0.4747
##
## Coefficients:
                       Estimate Std. Error t value Pr(>|t|)
##
                                   0.00385
                                              3.35 0.00083 ***
## (Intercept)
                        0.01287
## treat2
                        0.16416
                                   0.00552
                                             29.76 < 2e-16 ***
## reportwealth2
                        0.95537
                                   0.00665
                                           143.73 < 2e-16 ***
## treat2:reportwealth2 -0.02087
                                   0.00944
                                             -2.21 0.02701 *
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.14 on 9996 degrees of freedom
## Multiple R-squared: 0.922, Adjusted R-squared: 0.922
## F-statistic: 3.95e+04 on 3 and 9996 DF, p-value: <2e-16
##
## [[3]]
##
## Call:
## lm(formula = formula, data = sim_df)
##
## Residuals:
               1Q Median
      Min
                               3Q
                                      Max
## -0.4970 -0.0923 -0.0135 0.0697 0.9731
##
## Coefficients:
                       Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                       -0.00345
                                   0.00572
                                              -0.6
                                                        0.55
## treat3
                        0.33676
                                   0.00665
                                               50.7
                                                      <2e-16 ***
## reportwealth3
                        1.00270
                                   0.00903
                                             111.1
                                                      <2e-16 ***
## treat3:reportwealth3 0.16784
                                   0.01104
                                              15.2
                                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.146 on 9996 degrees of freedom
```

```
## Multiple R-squared: 0.917, Adjusted R-squared: 0.917
## F-statistic: 3.69e+04 on 3 and 9996 DF, p-value: <2e-16
##
##
## [[4]]
##
## lm(formula = formula, data = sim_df)
## Residuals:
       Min
                1Q Median
                               3Q
                                      Max
## -0.5054 -0.1138 -0.0254 0.0828 0.7491
## Coefficients:
##
                        Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                        0.15421
                                   0.00489
                                              31.5 <2e-16 ***
## treat4
                        0.18184
                                   0.00661
                                              27.5
                                                     <2e-16 ***
                                   0.00926
                                              93.0
## reportwealth4
                        0.86137
                                                     <2e-16 ***
## treat4:reportwealth4 0.28481
                                   0.01217
                                              23.4 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.167 on 9996 degrees of freedom
## Multiple R-squared: 0.891, Adjusted R-squared: 0.891
## F-statistic: 2.72e+04 on 3 and 9996 DF, p-value: <2e-16
lapply(1:4, function(i){
  # Generate formula
  y = paste0("new_wealth", i)
  x = paste0("reportwealth", i)
  formula = paste(y, x, sep = "~")
  # Run rdd
  RDestimate(formula, sim df)
})
f)
## [[1]]
## RDestimate(formula = formula, data = sim_df)
##
## Coefficients:
##
       LATE
              Half-BW Double-BW
##
      -0.192
                -0.186
                           -0.197
##
##
## [[2]]
##
## RDestimate(formula = formula, data = sim_df)
##
```

```
## Coefficients:
##
      LATE Half-BW Double-BW
     -0.206 -0.192 -0.203
##
##
## [[3]]
##
## Call:
## RDestimate(formula = formula, data = sim_df)
## Coefficients:
     LATE Half-BW Double-BW -0.302 -0.288 -0.316
##
##
##
##
## [[4]]
##
## Call:
## RDestimate(formula = formula, data = sim_df)
## Coefficients:
##
      LATE
             Half-BW Double-BW
## -0.200
             -0.202
                       -0.193
```