

HW6

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#HW6

```
library(estimatr)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.1
## v ggplot2    3.5.1      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.1
## v purrr      1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

#Question 1

```
load("mroz.RData")
```

```
mroz <- data
```

#a

```
lpm_model <- lm_robust(inlf ~ educ + exper + expersq, data = mroz)
summary(lpm_model)
```

```
##
## Call:
## lm_robust(formula = inlf ~ educ + exper + expersq, data = mroz)
##
## Standard error type: HC2
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper DF
## (Intercept) -0.1616014  0.0870177  -1.857 6.369e-02 -0.332429  0.0092263 749
## educ         0.0321910  0.0070564   4.562 5.919e-06  0.018338  0.0460436 749
## exper        0.0465279  0.0056923   8.174 1.273e-15  0.035353  0.0577027 749
## expersq      -0.0008996  0.0001805  -4.985 7.717e-07 -0.001254 -0.0005453 749
##
## Multiple R-squared:  0.169 , Adjusted R-squared:  0.1657
## F-statistic: 63.88 on 3 and 749 DF,  p-value: < 2.2e-16
```

```
#delta_p_inlf = -0.16 + .032 * delta_educ
```

```
#b
```

```
mroz$inflhat_lpm <- fitted(lpm_model)
summary(mroz$inflhat_lpm)
```

```
##      Min.    1st Qu.    Median      Mean   3rd Qu.      Max.
## -0.0006465  0.4205597  0.5954567  0.5683931  0.7315301  0.9841544
```

The range is -.00065 to .98415. This is a problem because the value going below 0 is nonsensical.

```
#c
```

```
probit_model <- glm(inlf ~ educ + exper + expersq, mroz, family = binomial(link = "probit"))
summary(probit_model)
```

```
##
## Call:
## glm(formula = inlf ~ educ + exper + expersq, family = binomial(link = "probit"),
##      data = mroz)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.9254937  0.2886193  -6.671 2.53e-11 ***
## educ         0.0971238  0.0221693   4.381 1.18e-05 ***
## exper        0.1271342  0.0177861   7.148 8.81e-13 ***
## expersq      -0.0023927  0.0005763  -4.152 3.29e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1029.75  on 752  degrees of freedom
## Residual deviance:  894.33  on 749  degrees of freedom
## AIC: 902.33
##
## Number of Fisher Scoring iterations: 4
```

```
mroz$inflhat_probit <- fitted(probit_model)
summary(mroz$inflhat_probit)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.07495 0.41844 0.59941 0.56871 0.74410 0.92065
```

The range is .075 to .921, which is not a problem because it is between 0 and 1.

```
#d
```

They are roughly the same statistical significance but their signs are different. This is what I expected because they are measuring the same effect but the magnitude of the probit model coefficient is not easily interpretable, while the coefficient for the LPM model is.

```

#e
beta_educ <- coef(probit_model)["educ"]
linear_preds <- probit_model$linear.predictors

marginal_effects_educ <- dnorm(linear_preds) * beta_educ
APE_educ <- mean(marginal_effects_educ)
APE_educ

```

```
## [1] 0.03278275
```

```

#f
betas <- coef(probit_model)
X <- model.matrix(probit_model)[,-1]
xbar <- colMeans(X)

linear_pred_avg <- betas[1] + sum(betas[-1] * xbar)

density_at_avg <- dnorm(linear_pred_avg)

PEA_educ <- density_at_avg * betas["educ"]
PEA_educ

```

```

## (Intercept)
## 0.03802895

```

#g

They are all roughly same except for the coefficient in the probit regression. This is what I would expect because LPM, APE, and PEA all are measuring very similar effects, while the magnitude of the coefficient in the probit model is not really interpretable.

#Question 3

```

load("kielmc.RData")

kielmc <- data

#a
only_1981 <- kielmc |>
  filter(year == 1981)

simple_model_81 <- lm_robust(lrprice ~ nearinc, only_1981)
summary(simple_model_81)

```

```

##
## Call:
## lm_robust(formula = lrprice ~ nearinc, data = only_1981)
##
## Standard error type: HC2
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper DF

```

```
## (Intercept) 11.4785    0.03170 362.102 6.145e-210 11.4158    11.541 140
## nearinc     -0.4026    0.07159  -5.623 9.829e-08  -0.5441    -0.261 140
##
## Multiple R-squared:  0.2172 ,    Adjusted R-squared:  0.2116
## F-statistic: 31.62 on 1 and 140 DF,  p-value: 9.829e-08
```

The results show that the incinerator had an adverse impact on home prices. However, this model does not contain any control variable to make causality a more reasonable assumption.

```
#b
model_w_controls_81 <- lm_robust(lprice ~ nearinc + age + agesq + rooms + baths + lintst + larea + lland, data = only_1981)
summary(model_w_controls_81)
```

```
##
## Call:
## lm_robust(formula = lprice ~ nearinc + age + agesq + rooms +
##          baths + lintst + larea + lland, data = only_1981)
##
## Standard error type:  HC2
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)    CI Lower    CI Upper  DF
## (Intercept)  8.674e+00  0.7347586 11.8054 1.906e-22  7.221e+00  1.013e+01 133
## nearinc     -1.364e-01  0.0818480 -1.6665 9.797e-02 -2.983e-01  2.549e-02 133
## age        -8.317e-03  0.0026353 -3.1562 1.978e-03 -1.353e-02 -3.105e-03 133
## agesq       4.145e-05  0.0000227  1.8265 7.002e-02 -3.438e-06  8.634e-05 133
## rooms       1.760e-02  0.0259984  0.6771 4.995e-01 -3.382e-02  6.903e-02 133
## baths       1.315e-01  0.0470726  2.7943 5.971e-03  3.843e-02  2.246e-01 133
## lintst      -8.171e-02  0.0614364 -1.3300 1.858e-01 -2.032e-01  3.981e-02 133
## larea       3.385e-01  0.0955037  3.5446 5.434e-04  1.496e-01  5.274e-01 133
## lland       8.070e-02  0.0401501  2.0099 4.647e-02  1.282e-03  1.601e-01 133
##
## Multiple R-squared:  0.7619 ,    Adjusted R-squared:  0.7475
## F-statistic: 74.72 on 8 and 133 DF,  p-value: < 2.2e-16
```

The coefficient on nearinc is smaller and the standard error is larger. This goes a long way towards resolving the issues found in a) but we still can't be sure there aren't other variables we need to control for to ensure causality.

```
#c
only_1978 <- kielmc |>
  filter(year == 1978)

simple_model_78 <- lm_robust(lrprice ~ nearinc, only_1978)
summary(simple_model_78)
```

```
##
## Call:
## lm_robust(formula = lrprice ~ nearinc, data = only_1978)
##
## Standard error type:  HC2
##
```

```
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper DF
## (Intercept)  11.2854    0.02510 449.679 1.336e-272  11.2359  11.3350 177
## nearinc      -0.3399    0.06245  -5.443 1.729e-07  -0.4632  -0.2167 177
##
## Multiple R-squared:  0.1855 , Adjusted R-squared:  0.1809
## F-statistic: 29.62 on 1 and 177 DF, p-value: 1.729e-07
```

The coefficient for nearinc is almost as large as it was when using the 1981 data. This tells us that there is something else about the areas where the incinerators were placed that is hurting home prices and that is what our analysis in a) is picking up because the incinerators weren't even built in 1978.

```
#d
kielmc_81ind <- kielmc |>
  mutate(y81 = ifelse(year == 1981, 1, 0))

pooled_model <- lm_robust(lrprice ~ nearinc + y81 + y81 * nearinc, kielmc_81ind)
summary(pooled_model)
```

```
##
## Call:
## lm_robust(formula = lrprice ~ nearinc + y81 + y81 * nearinc,
##           data = kielmc_81ind)
##
## Standard error type: HC2
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper DF
## (Intercept)  11.28542    0.02510 449.6786 0.000e+00  11.2360  11.3348 317
## nearinc      -0.33992    0.06245  -5.4428 1.052e-07  -0.4628  -0.2170 317
## y81           0.19309    0.04043   4.7758 2.740e-06   0.1135   0.2726 317
## nearinc:y81  -0.06265    0.09501  -0.6594 5.101e-01  -0.2496   0.1243 317
##
## Multiple R-squared:  0.246 , Adjusted R-squared:  0.2388
## F-statistic: 27.55 on 3 and 317 DF, p-value: 7.423e-16
```

This model tells us that for home prices in 1981, being near the incinerator on average reduced the price by $(.4628 + .2496 = .7124)\%$. But, about 2/3 of the effect on home prices comes regardless of whether they were measured before or after the incinerator was built. So, while the incinerator does have some impact on housing prices, there is something else area having a larger impact.

```
#e
pooled_model_control <- lm_robust(lrprice ~ nearinc + y81 + y81 * nearinc + age + agesq + rooms + baths +
summary(pooled_model_control)
```

```
##
## Call:
## lm_robust(formula = lrprice ~ nearinc + y81 + y81 * nearinc +
##           age + agesq + rooms + baths + lintst + larea + lland, data = kielmc_81ind)
##
## Standard error type: HC2
##
```

```
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|) CI Lower CI Upper DF
## (Intercept)  7.652e+00  5.375e-01  14.235 9.948e-36  6.594e+00  8.709460 310
## nearinc      3.223e-02  6.485e-02   0.497 6.195e-01 -9.537e-02  0.159829 310
## y81          4.260e-01  2.720e-02  15.660 3.969e-41  3.725e-01  0.479497 310
## age         -8.359e-03  1.657e-03  -5.045 7.721e-07 -1.162e-02 -0.005099 310
## agesq        3.763e-05  1.137e-05   3.310 1.042e-03  1.526e-05  0.000060 310
## rooms        4.733e-02  1.776e-02   2.665 8.095e-03  1.239e-02  0.082279 310
## baths        9.428e-02  2.846e-02   3.312 1.036e-03  3.827e-02  0.150285 310
## lintst       -6.145e-02  3.569e-02  -1.722 8.616e-02 -1.317e-01  0.008786 310
## larea        3.508e-01  6.295e-02   5.572 5.461e-08  2.269e-01  0.474632 310
## lland         9.985e-02  3.410e-02   2.928 3.663e-03  3.275e-02  0.166939 310
## nearinc:y81 -1.315e-01  6.082e-02  -2.162 3.135e-02 -2.512e-01 -0.011844 310
##
## Multiple R-squared:  0.7904 , Adjusted R-squared:  0.7837
## F-statistic: 135 on 10 and 310 DF, p-value: < 2.2e-16
```

#Yes, it makes sense to include these control variables because they are all plausibly have meaningful impacts on home price and could vary depending on whether the home is close to the incinerator or not.

```
#f
nearinc_data <- kielmc_81ind |>
  filter(nearinc == 1)

summary(nearinc_data)
```

```
##           year           age           agesq           nbh
## Min.      :1978   Min.      : 0.00   Min.      : 0.0   Min.      :0.000
## 1st Qu.:1978   1st Qu.: 11.75   1st Qu.: 138.2   1st Qu.:0.750
## Median :1978   Median : 25.50   Median : 650.5   Median :4.000
## Mean      :1979   Mean      : 34.85   Mean      : 2347.3   Mean      :2.625
## 3rd Qu.:1981   3rd Qu.: 51.75   3rd Qu.: 2679.8   3rd Qu.:4.000
## Max.      :1981   Max.      :189.00   Max.      :35721.0   Max.      :4.000
##           cbd           intst           lintst           price
## Min.      : 1000   Min.      : 1000   Min.      :6.908   Min.      : 31000
## 1st Qu.: 3000   1st Qu.: 4000   1st Qu.:8.294   1st Qu.: 48000
## Median : 4000   Median : 6000   Median :8.700   Median : 62000
## Mean      : 5458   Mean      : 6521   Mean      :8.567   Mean      : 75465
## 3rd Qu.: 6250   3rd Qu.: 7000   3rd Qu.:8.854   3rd Qu.: 83750
## Max.      :14000   Max.      :17000   Max.      :9.741   Max.      :300000
##           rooms           area           land           baths
## Min.      :4.000   Min.      : 750   Min.      : 1710   Min.      :1.000
## 1st Qu.:5.000   1st Qu.:1394   1st Qu.: 8253   1st Qu.:1.000
## Median :6.000   Median :1651   Median : 12500   Median :2.000
## Mean      :6.083   Mean      :1888   Mean      : 22392   Mean      :1.844
## 3rd Qu.:7.000   3rd Qu.:2228   3rd Qu.: 21290   3rd Qu.:2.000
## Max.      :9.000   Max.      :5136   Max.      :282704   Max.      :4.000
##           dist           ldist           wind           lprice
## Min.      : 5000   Min.      :8.517   Min.      : 3.000   Min.      :10.34
## 1st Qu.: 8500   1st Qu.:9.048   1st Qu.: 3.000   1st Qu.:10.78
## Median :10850   Median :9.292   Median : 5.000   Median :11.03
## Mean      :10539   Mean      :9.223   Mean      : 5.177   Mean      :11.11
## 3rd Qu.:12325   3rd Qu.:9.419   3rd Qu.: 5.000   3rd Qu.:11.34
```

```
## Max. :15600 Max. :9.655 Max. :11.000 Max. :12.61
## y81 larea lland y81ldist
## Min. :0.0000 Min. :6.620 Min. : 7.444 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:7.240 1st Qu.: 9.018 1st Qu.:0.000
## Median :0.0000 Median :7.409 Median : 9.433 Median :0.000
## Mean :0.4167 Mean :7.461 Mean : 9.552 Mean :3.839
## 3rd Qu.:1.0000 3rd Qu.:7.709 3rd Qu.: 9.966 3rd Qu.:9.205
## Max. :1.0000 Max. :8.544 Max. :12.552 Max. :9.616
## lintstsq nearinc y81nrinc rprice lrprice
## Min. :47.72 Min. :1 Min. :0.0000 Min. : 31000 Min. :10.34
## 1st Qu.:68.79 1st Qu.:1 1st Qu.:0.0000 1st Qu.: 47464 1st Qu.:10.77
## Median :75.68 Median :1 Median :0.0000 Median : 54500 Median :10.91
## Mean :73.88 Mean :1 Mean :0.4167 Mean : 66579 Mean :11.00
## 3rd Qu.:78.39 3rd Qu.:1 3rd Qu.:1.0000 3rd Qu.: 70675 3rd Qu.:11.17
## Max. :94.89 Max. :1 Max. :1.0000 Max. :300000 Max. :12.61
```

```
farinc_data <- kielmc_81ind |>
  filter(nearinc == 0)

summary(farinc_data)
```

```
## year age agesq nbh
## Min. :1978 Min. : 0.00 Min. : 0.0 Min. :0.000
## 1st Qu.:1978 1st Qu.: 0.00 1st Qu.: 0.0 1st Qu.:0.000
## Median :1978 Median : 1.00 Median : 1.0 Median :2.000
## Mean :1979 Mean : 10.82 Mean : 969.5 Mean :2.031
## 3rd Qu.:1981 3rd Qu.: 8.00 3rd Qu.: 64.0 3rd Qu.:5.000
## Max. :1981 Max. :188.00 Max. :35344.0 Max. :6.000
## cbd intst lintst price
## Min. : 9000 Min. : 7000 Min. : 8.854 Min. : 26000
## 1st Qu.:14000 1st Qu.:14000 1st Qu.: 9.547 1st Qu.: 76900
## Median :21000 Median :22000 Median : 9.999 Median : 94376
## Mean :20244 Mean :20676 Mean : 9.870 Mean :104905
## 3rd Qu.:24000 3rd Qu.:26000 3rd Qu.:10.166 3rd Qu.:129900
## Max. :35000 Max. :34000 Max. :10.434 Max. :234552
## rooms area land baths dist
## Min. : 4.0 Min. : 735 Min. : 7500 Min. :1.000 Min. :16000
## 1st Qu.: 6.0 1st Qu.:1836 1st Qu.: 37026 1st Qu.:2.000 1st Qu.:19000
## Median : 7.0 Median :2240 Median : 44001 Median :3.000 Median :25500
## Mean : 6.8 Mean :2200 Mean : 46985 Mean :2.551 Mean :25058
## 3rd Qu.: 7.0 3rd Qu.:2580 3rd Qu.: 47480 3rd Qu.:3.000 3rd Qu.:28600
## Max. :10.0 Max. :4056 Max. :544500 Max. :4.000 Max. :40000
## ldist wind lprice y81
## Min. : 9.680 Min. : 3.000 Min. :10.17 Min. :0.0000
## 1st Qu.: 9.852 1st Qu.: 7.000 1st Qu.:11.25 1st Qu.:0.0000
## Median :10.146 Median : 7.000 Median :11.46 Median :0.0000
## Mean :10.100 Mean : 7.747 Mean :11.49 Mean :0.4533
## 3rd Qu.:10.261 3rd Qu.:11.000 3rd Qu.:11.77 3rd Qu.:1.0000
## Max. :10.597 Max. :11.000 Max. :12.37 Max. :1.0000
## larea lland y81ldist lintstsq nearinc
## Min. :6.600 Min. : 8.923 Min. : 0.000 Min. : 78.39 Min. :0
## 1st Qu.:7.515 1st Qu.:10.519 1st Qu.: 0.000 1st Qu.: 91.14 1st Qu.:0
## Median :7.714 Median :10.692 Median : 0.000 Median : 99.98 Median :0
## Mean :7.655 Mean :10.622 Mean : 4.557 Mean : 97.57 Mean :0
```

```
## 3rd Qu.:7.856 3rd Qu.:10.768 3rd Qu.:10.043 3rd Qu.:103.35 3rd Qu.:0
## Max. :8.308 Max. :13.208 Max. :10.569 Max. :108.87 Max. :0
## y81nrinc rprice lrprice
## Min. :0 Min. : 26000 Min. :10.17
## 1st Qu.:0 1st Qu.: 73000 1st Qu.:11.20
## Median :0 Median : 89500 Median :11.40
## Mean :0 Mean : 91035 Mean :11.37
## 3rd Qu.:0 3rd Qu.:107527 3rd Qu.:11.59
## Max. :0 Max. :180147 Max. :12.10
```

This person does have a point as the houses near the incinerator appear to have very different characteristics than the houses away from the incinerator. To provide an unbiased estimate we'd need to know that the prices of the homes near and not near to the incinerator would have had a similar trend. Data on home prices in the years prior to 1978 would tell us if our assumption is reasonable.

#g

This would reduce the internal validity of the research design because the closing of the schools would also negatively effect the price of homes in the area, so we'd be unable to determine if the school closing or the incinerator would be responsible for the change in price.

Question 2

a)

The coefficient estimates are consistent and unbiased, so they get closer to the value of the true parameter as sample size increases. The standard errors shrink as the sample size increases.

b)

See handwritten answer

c)

The estimate would be less precise because we would have less variation in income to inform our estimate.

d)

It would improve the precision of the estimates because education is likely to have a strong relationship with whether or not someone voted.

HW6

$$2. b) se(\hat{\beta}_j) = \frac{\sqrt{\sum \hat{u}_i^2 / (n - k - 1)}}{\sqrt{SST_j (1 - R_j^2)}}$$