

ARE 213 Problem Set 3

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Import Data

```
#setwd("~/Dropbox/Berkeley_tings/Fall 2018/ARE213/Problem Sets/SharedFiles/are213/PS2b")
setwd("C:\\Users\\will-\\Desktop\\are213\\PS3")
#setwd("C:\\Users\\Will\\Desktop\\are213\\PS3")
#setwd("C:\\Users\\Peter\\Documents\\Berkeley\\Courses\\ARE 213\\Problem Sets\\3")
site_covariates <- read.dta("sitecovariates.dta")
all_sites <- read.dta13("allsites.dta")
all_covariates <- read.dta("allcovariates.dta")
mile_data <- read.dta("2miledata.dta")
```

1a - tract housing analysis

Outcome Variable: Median housing price in 2000 Control Variables: NPL site in tract in 2000, mean housing price in 1980

```
ols1 <- felm(lnmdvalhs0 ~ npl2000 + lnmeanhs8, data=all_sites)
summary(ols1, robust=T)
```

```
##
## Call:
##   felm(formula = lnmdvalhs0 ~ npl2000 + lnmeanhs8, data = all_sites)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.0581 -0.2173 -0.0241  0.1921  2.8403
##
## Coefficients:
##              Estimate Robust s.e t value Pr(>|t|)
## (Intercept)   2.40427     0.12410  19.373 < 2e-16 ***
## npl2000        0.04004     0.01195   3.352 0.000804 ***
## lnmeanhs8     0.85575     0.01147  74.622 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4057 on 42971 degrees of freedom
## Multiple R-squared(full model): 0.5792   Adjusted R-squared: 0.5791
## Multiple R-squared(proj model): 0.5792   Adjusted R-squared: 0.5791
## F-statistic(full model, *iid*):2.957e+04 on 2 and 42971 DF, p-value: < 2.2e-16
## F-statistic(proj model): 2784 on 2 and 42971 DF, p-value: < 2.2e-16
```

Outcome Variable: Median housing price in 2000 Control Variables: NPL site in tract in 2000, mean housing price in 1980 and various housing characteristics

```
ols2 <- felm(lnmdvalhs0 ~ npl2000 + lnmeanhs8 + firestoveheat80 + noaircond80 +
  nofullkitchen80 + zerofullbath80 + bedrms0_80occ + bedrms1_80occ +
  bedrms2_80occ + bedrms3_80occ +bedrms4_80occ + blt0_1yrs80occ +
```

```

      blt2_5yrs80occ +blt6_10yrs80occ +blt10_20yrs80occ +blt20_30yrs80occ +
      blt30_40yrs80occ +detach80occ+attach80occ, data=all_sites)
summary(ols2, robust=T)

```

```

##
## Call:
##   felm(formula = lnmdvalhs0 ~ npl2000 + lnmeanhs8 + firestoveheat80 +      noaircond80 + nofullkitc
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.9335 -0.1848 -0.0048  0.1896  2.8751
##
## Coefficients:
##              Estimate Robust s.e t value Pr(>|t|)
## (Intercept)    3.170e+00  2.142e-01  14.796 < 2e-16 ***
## npl2000         4.395e-02  1.156e-02   3.801 0.000145 ***
## lnmeanhs8       8.605e-01  1.791e-02  48.054 < 2e-16 ***
## firestoveheat80 -2.917e-02  1.964e-02  -1.485 0.137459
## noaircond80      3.201e-01  7.340e-03  43.618 < 2e-16 ***
## nofullkitchen80 -1.683e+00  1.680e-01 -10.015 < 2e-16 ***
## zerofullbath80   6.795e-01  1.350e-01   5.033 4.85e-07 ***
## bedrms0_80occ    -6.831e-01  2.224e-01  -3.071 0.002133 **
## bedrms1_80occ    -3.823e-01  7.846e-02  -4.873 1.11e-06 ***
## bedrms2_80occ    -1.008e+00  6.532e-02 -15.427 < 2e-16 ***
## bedrms3_80occ    -1.065e+00  6.664e-02 -15.989 < 2e-16 ***
## bedrms4_80occ    -6.188e-01  7.197e-02  -8.599 < 2e-16 ***
## blt0_1yrs80occ   -1.365e-01  4.249e-02  -3.213 0.001315 **
## blt2_5yrs80occ   -1.633e-01  2.985e-02  -5.472 4.46e-08 ***
## blt6_10yrs80occ  -1.240e-01  2.245e-02  -5.522 3.37e-08 ***
## blt10_20yrs80occ -6.766e-02  1.822e-02  -3.713 0.000205 ***
## blt20_30yrs80occ  7.271e-04  1.903e-02   0.038 0.969522
## blt30_40yrs80occ -6.922e-02  2.547e-02  -2.718 0.006578 **
## detach80occ      -8.333e-05  2.463e-02  -0.003 0.997301
## attach80occ       2.686e-01  2.749e-02   9.771 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3711 on 42954 degrees of freedom
## Multiple R-squared(full model): 0.648   Adjusted R-squared: 0.6478
## Multiple R-squared(proj model): 0.648   Adjusted R-squared: 0.6478
## F-statistic(full model, *iid*): 4162 on 19 and 42954 DF, p-value: < 2.2e-16
## F-statistic(proj model): 1871 on 19 and 42954 DF, p-value: < 2.2e-16

```

Outcome Variable: Median housing price in 2000 Control Variables: NPL site in tract in 2000, mean housing price in 1980 and various housing and demographic characteristics

```

ols3 <- felm(lnmdvalhs0 ~ npl2000 + lnmeanhs8 + firestoveheat80 + noaircond80 +
  nofullkitchen80 + zerofullbath80 + bedrms0_80occ + bedrms1_80occ +
  bedrms2_80occ + bedrms3_80occ +bedrms4_80occ + blt0_1yrs80occ +
  blt2_5yrs80occ +blt6_10yrs80occ +blt10_20yrs80occ +blt20_30yrs80occ +
  blt30_40yrs80occ +detach80occ+attach80occ+pop_den8+shrbld8+
  shrhsp8+child8+old8+shrford8+ffh8+smhse8+hsdrop8+no_hs_diploma8+
  ba_or_better8+unemp8+povrat8+welfare8+avhhin8+tothsun8+ownocc8, data=all_sites)
summary(ols3, robust=T)

```

```
##
## Call:
##   felm(formula = lnmdvalhs0 ~ npl2000 + lnmeanhs8 + firestoveheat80 +      noaircond80 + nofullkitch
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.8489 -0.1768  0.0039  0.1822  2.2764
##
## Coefficients:
##              Estimate Robust s.e t value Pr(>|t|)
## (Intercept)    5.892e+00  2.200e-01  26.783 < 2e-16 ***
## npl2000         7.392e-02  9.693e-03   7.626 2.47e-14 ***
## lnmeanhs8       5.584e-01  1.945e-02  28.708 < 2e-16 ***
## firestoveheat80 -2.632e-02  2.053e-02  -1.282 0.199693
## noaircond80      4.374e-01  8.811e-03  49.635 < 2e-16 ***
## nofullkitchen80 -6.013e-01  1.535e-01  -3.917 8.96e-05 ***
## zerofullbath80   5.089e-01  1.206e-01   4.221 2.44e-05 ***
## bedrms0_80occ    -7.461e-01  2.248e-01  -3.319 0.000906 ***
## bedrms1_80occ    -2.029e-01  7.749e-02  -2.619 0.008830 **
## bedrms2_80occ    -4.166e-01  5.806e-02  -7.176 7.31e-13 ***
## bedrms3_80occ    -5.544e-01  5.583e-02  -9.930 < 2e-16 ***
## bedrms4_80occ    -4.961e-01  6.673e-02  -7.435 1.07e-13 ***
## blt0_1yrs80occ   9.559e-02  4.351e-02   2.197 0.028019 *
## blt2_5yrs80occ   1.633e-01  2.814e-02   5.802 6.62e-09 ***
## blt6_10yrs80occ  1.168e-01  2.153e-02   5.426 5.80e-08 ***
## blt10_20yrs80occ 6.760e-02  1.556e-02   4.343 1.41e-05 ***
## blt20_30yrs80occ 1.914e-02  1.446e-02   1.323 0.185760
## blt30_40yrs80occ 3.502e-02  2.458e-02   1.425 0.154253
## detach80occ      -2.706e-01  2.282e-02 -11.860 < 2e-16 ***
## attach80occ      -2.643e-01  2.630e-02 -10.049 < 2e-16 ***
## pop_den8         7.436e-06  3.903e-07  19.054 < 2e-16 ***
## shrblk8          -1.129e-01  1.311e-02  -8.609 < 2e-16 ***
## shrhsp8          -2.854e-01  1.975e-02 -14.448 < 2e-16 ***
## child8           -5.629e-01  5.529e-02 -10.180 < 2e-16 ***
## old8             -3.520e-01  4.442e-02  -7.925 2.34e-15 ***
## shrfor8          1.230e+00  4.333e-02  28.384 < 2e-16 ***
## ffh8             -4.794e-02  3.423e-02  -1.400 0.161391
## smhse8           4.376e-01  2.260e-02  19.369 < 2e-16 ***
## hsdrop8          1.450e-02  2.438e-02   0.595 0.552085
## no_hs_diploma8    -3.103e-01  3.401e-02  -9.124 < 2e-16 ***
## ba_or_better8     4.784e-01  3.335e-02  14.345 < 2e-16 ***
## unemp8           -1.236e+00  7.165e-02 -17.243 < 2e-16 ***
## povrat8          -3.838e-01  4.938e-02  -7.773 7.84e-15 ***
## welfare8         8.570e-01  6.608e-02  12.970 < 2e-16 ***
## avhhin8          1.306e-05  6.865e-07  19.030 < 2e-16 ***
## tothsun8         1.059e-05  6.411e-06   1.651 0.098732 .
## ownocc8          -1.325e-04  1.017e-05 -13.024 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3244 on 42937 degrees of freedom
## Multiple R-squared(full model): 0.7311   Adjusted R-squared: 0.7309
## Multiple R-squared(proj model): 0.7311   Adjusted R-squared: 0.7309
## F-statistic(full model, *iid*): 3242 on 36 and 42937 DF, p-value: < 2.2e-16
```

```
## F-statistic(proj model): 2641 on 36 and 42937 DF, p-value: < 2.2e-16
```

Outcome Variable: Median housing price in 2000 Control Variables: NPL site in tract in 2000, mean housing price in 1980 and various housing and demographic characteristics and state fixed effects (excluded variables that returned errors in fixed effects treatment bedrooms_80occ, blt40_yrs80occ, percent mobile home single family housing 1980)

```
ols4 <- felm(lnmdvalhs0 ~ npl2000 + lnmeanhs8 + firestoveheat80 + noaircond80 +
  nofullkitchen80 + zerofullbath80 + bedrooms_80occ + bedrooms1_80occ +
  bedrooms2_80occ + bedrooms3_80occ + bedrooms4_80occ + blt0_1yrs80occ +
  blt2_5yrs80occ + blt6_10yrs80occ + blt10_20yrs80occ + blt20_30yrs80occ +
  blt30_40yrs80occ + detach80occ + attach80occ + pop_den8 + shrblk8 +
  shrhsp8 + child8 + old8 + shrfor8 + ffh8 + smhse8 + hsdrop8 + no_hs_diploma8 +
  ba_or_better8 + unemp8 + povrat8 + welfare8 + avhhin8 + tothsun8 +
  ownocc8 | statefips, data=all_sites)
summary(ols4, robust=T)
```

```
##
```

```
## Call:
```

```
##      felm(formula = lnmdvalhs0 ~ npl2000 + lnmeanhs8 + firestoveheat80 +      noaircond80 + nofullkitch
```

```
##
```

```
## Residuals:
```

```
##      Min      1Q   Median      3Q      Max
## -3.6913 -0.1541  0.0003  0.1584  2.5369
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Robust s.e t value Pr(>|t|)
## npl2000          6.635e-02  8.797e-03   7.542 4.71e-14 ***
## lnmeanhs8        4.964e-01  2.067e-02  24.010 < 2e-16 ***
## firestoveheat80   7.965e-02  3.192e-02   2.495 0.012599 *
## noaircond80       3.252e-01  9.485e-03  34.282 < 2e-16 ***
## nofullkitchen80  -5.408e-01  1.463e-01  -3.697 0.000218 ***
## zerofullbath80    5.940e-01  1.142e-01   5.202 1.98e-07 ***
## bedrooms_80occ   -4.740e-01  2.239e-01  -2.117 0.034296 *
## bedrooms1_80occ  -1.241e-01  7.453e-02  -1.666 0.095801 .
## bedrooms2_80occ  -3.759e-01  5.548e-02  -6.775 1.26e-11 ***
## bedrooms3_80occ  -4.438e-01  5.368e-02  -8.267 < 2e-16 ***
## bedrooms4_80occ  -4.767e-01  6.305e-02  -7.561 4.06e-14 ***
## blt0_1yrs80occ    1.017e-01  4.365e-02   2.330 0.019821 *
## blt2_5yrs80occ    1.129e-01  2.625e-02   4.301 1.70e-05 ***
## blt6_10yrs80occ   4.059e-02  2.081e-02   1.951 0.051089 .
## blt10_20yrs80occ -2.183e-02  1.409e-02  -1.549 0.121327
## blt20_30yrs80occ -2.801e-02  1.308e-02  -2.142 0.032204 *
## blt30_40yrs80occ  1.059e-02  2.250e-02   0.470 0.638055
## detach80occ      -2.530e-01  2.210e-02 -11.448 < 2e-16 ***
## attach80occ      -1.886e-01  2.541e-02  -7.421 1.19e-13 ***
## pop_den8         6.770e-06  3.880e-07  17.446 < 2e-16 ***
## shrblk8          -9.578e-02  1.265e-02  -7.571 3.78e-14 ***
## shrhsp8          -8.938e-02  2.124e-02  -4.208 2.58e-05 ***
## child8           -3.631e-01  5.189e-02  -6.997 2.65e-12 ***
## old8             -1.940e-01  4.325e-02  -4.486 7.26e-06 ***
## shrfor8          5.863e-01  3.925e-02  14.937 < 2e-16 ***
## ffh8             -1.125e-01  3.324e-02  -3.384 0.000716 ***
## smhse8           3.747e-01  2.290e-02  16.364 < 2e-16 ***
## hsdrop8          2.231e-02  2.323e-02   0.961 0.336699
```

```
## no_hs_diploma8      -3.568e-01  3.294e-02 -10.829 < 2e-16 ***
## ba_or_better8       5.054e-01  3.464e-02  14.590 < 2e-16 ***
## unemprt8            -1.377e+00  7.431e-02 -18.527 < 2e-16 ***
## povrat8             -9.174e-02  4.770e-02  -1.923 0.054441 .
## welfare8            2.959e-01  6.490e-02   4.559 5.14e-06 ***
## avhhin8             1.383e-05  6.198e-07  22.314 < 2e-16 ***
## tothsun8            1.429e-05  6.148e-06   2.324 0.020119 *
## ownocc8             -1.323e-04  9.683e-06 -13.667 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2941 on 42887 degrees of freedom
## Multiple R-squared(full model): 0.7793   Adjusted R-squared: 0.7789
## Multiple R-squared(proj model): 0.6876   Adjusted R-squared: 0.687
## F-statistic(full model, *iid*): 1761 on 86 and 42887 DF, p-value: < 2.2e-16
## F-statistic(proj model): 2032 on 36 and 42887 DF, p-value: < 2.2e-16
```

The coefficient for NPL status is significant and positive in all of the regressions. Since all of the coefficients for NPL are positive, it suggests this designation has an unambiguously positive correlation with housing value. The magnitude of the coefficient on NPL term doesn't change notably as the additional demographic and economic variables are added .

In the first three regressions, we have to make the assumptions associated with a selection on observables research design.

The coefficient will be unbiased if the following assumptions are met: - residuals are mean zero - residuals are homoskedastic with finite variance - residuals are uncorrelated with NPL2000

The fixed effect design, on the other hand, does not have to rely on the selection of observables assumption but rather that the the observables individual specific effects are not correlated with the error term and that the observeables are not correlated with the individual specific effects.

1b - treatment vs. comparison

```
group.ttest = function(x, group = as.factor(all_covariates$npl2000 == 1) ){
  return(
    unlist(
      t.test( x ~ group)[c("estimate", "p.value")]
    )
  )
}
comp_cov = t(sapply(all_covariates[,c(1,3:ncol(all_covariates))], group.ttest))
colnames(comp_cov) <- c("non-npl2000", "npl2000", "p-value")
rownames(comp_cov) <- names(all_covariates[c(1,3:ncol(all_covariates))])
stargazer(comp_cov, title = "Table 1b1: difference in covariates", type = "text")
```

```
##
## Table 1b1: difference in covariates
## =====
##                non-npl2000    npl2000    p-value
## -----
## npl1990                0         0.757         0
## pop_den8             5,424.074    1,406.947         0
## shrblk8               0.117         0.091    0.0001
```

## shrhsp8	0.072	0.052	0.00000
## child8	0.279	0.294	0
## shrfor8	0.071	0.055	0
## ffh8	0.190	0.162	0
## smhse8	0.515	0.544	0
## hsdrop8	0.136	0.138	0.590
## no_hs_diploma8	0.315	0.343	0
## ba_or_better8	0.174	0.139	0
## unemprt8	0.066	0.069	0.006
## povrat8	0.113	0.106	0.017
## welfare8	0.076	0.074	0.192
## favinc8	24,236.770	22,757.300	0
## avhhin8	21,510.180	20,340.160	0.00000
## meanrnt80	281.980	264.639	0
## mdvalhs9	99,115.530	100,065.600	0.725
## meanrnt9	534.609	516.100	0.002
## mdvalhs0	150,718.500	139,040.900	0.00002
## meanrnt0	708.394	684.161	0.003
## tothsun8	1,347.862	1,391.997	0.032
## ownocc8	800.653	907.861	0
## owner_occupied80	0.619	0.682	0.0004
## bltlast5yrs80	0.154	0.143	0.005
## bltlast10yrs80	0.289	0.283	0.380
## firestoveheat80	0.043	0.054	0.0001
## noaircond80	0.427	0.490	0
## nofullkitchen80	0.017	0.019	0.077
## zerofullbath80	0.023	0.025	0.014
## northeast	0.224	0.380	0
## midwest	0.231	0.218	0.335
## south	0.314	0.236	0
## west	0.231	0.166	0.00000
## meanhs8	68,984.340	58,045.370	0.001
## bedrms02_80	0.468	0.448	0.0004
## bedrms34_80	0.505	0.525	0.0002
## detach80	0.622	0.677	0
## bedrms0_80occ	0.003	0.003	0.029
## bedrms1_80occ	0.047	0.044	0.043
## bedrms2_80occ	0.265	0.278	0.001
## bedrms3_80occ	0.480	0.484	0.192
## bedrms4_80occ	0.167	0.158	0.001
## bedrms5_80occ	0.038	0.033	0.00000
## blt0_1yrs80occ	0.038	0.035	0.003
## blt2_5yrs80occ	0.108	0.110	0.522
## blt6_10yrs80occ	0.119	0.132	0.0001
## blt10_20yrs80occ	0.186	0.194	0.036
## blt20_30yrs80occ	0.185	0.193	0.062
## blt30_40yrs80occ	0.103	0.106	0.384
## blt40_yrs80occ	0.261	0.230	0.00001
## detach80occ	0.879	0.876	0.590
## attach80occ	0.072	0.037	0
## mobile80occ	0.049	0.086	0
## occupied80	0.934	0.941	0.00000
## bltmore30_80	0.340	0.333	0.350
## nbr_dummy	0.112	0.441	0

```
## -----
group.ttest = function(x, group = as.factor(site_covariates$hrs_82 > 28.5) ){
  return(
    unlist(
      t.test( x ~ group)[c("estimate", "p.value")]
    )
  )
}
comp_cov = t(sapply(site_covariates[,c(1,3:58)], group.ttest))
colnames(comp_cov) <- c("Less than 28.5", "Greater than 28.5", "p-value")
rownames(comp_cov) <- names(site_covariates[c(1,3:58)])
stargazer(comp_cov, title = "Table 1b2: Group by 28.5 - Difference in covariates", type = "text")
```

```
##
## Table 1b2: Group by 28.5 - Difference in covariates
## =====
##                Less than 28.5 Greater than 28.5 p-value
## -----
## hrs_82                15.538                44.468                0
## npl2000                0.160                0.990                0
## pop_den8              1,670.241            1,157.464            0.068
## shrblk8                0.113                0.071            0.037
## shrhsp8                0.044                0.042            0.841
## child8                0.293                0.294            0.958
## shrfor8                0.053                0.051            0.735
## ffh8                  0.188                0.158            0.018
## smhse8                0.603                0.562            0.001
## hsdrop8               0.145                0.134            0.235
## no_hs_diploma8        0.405                0.343            0.00000
## ba_or_better8         0.100                0.138            0.00000
## unemprt8              0.086                0.071            0.001
## povrat8               0.114                0.100            0.109
## welfare8              0.089                0.074            0.041
## favinc8              21,657.670            23,041.600            0.005
## avhhin8              19,635.320            20,868.850            0.013
## meanrnt80             247.313                268.205            0.0004
## mdvalhs9             80,185.070            96,752.080            0.005
## meanrnt9              470.387                527.323            0.001
## mdvalhs0            115,479.000            135,435.600            0.001
## meanrnt0              609.843                690.646            0.0002
## tothsun8             1,356.740            1,352.810            0.951
## ownocc8              906.370                901.853            0.922
## owner_occupied80      0.679                0.680            0.959
## bltlast5yrs80         0.119                0.140            0.050
## bltlast10yrs80        0.237                0.281            0.012
## firestoveheat80       0.053                0.052            0.814
## noaircond80           0.506                0.480            0.253
## nofullkitchen80       0.021                0.019            0.402
## zerofullbath80        0.032                0.026            0.089
## northeast             0.331                0.477            0.001
## midwest               0.348                0.225            0.004
## south                 0.215                0.193            0.552
## west                  0.105                0.105            0.989
## meanhs8              45,027.340            52,136.800            0.0002
```

```
## bedrms02_80      0.469      0.444      0.108
## bedrms34_80      0.510      0.529      0.202
## detach80         0.658      0.690      0.181
## bedrms0_80occ    0.002      0.002      0.458
## bedrms1_80occ    0.046      0.044      0.617
## bedrms2_80occ    0.308      0.269      0.001
## bedrms3_80occ    0.483      0.488      0.636
## bedrms4_80occ    0.134      0.163      0.00005
## bedrms5_80occ    0.027      0.033      0.010
## blt0_1yrs80occ   0.027      0.033      0.042
## blt2_5yrs80occ   0.089      0.104      0.116
## blt6_10yrs80occ  0.109      0.130      0.018
## blt10_20yrs80occ 0.178      0.203      0.025
## blt20_30yrs80occ 0.185      0.193      0.520
## blt30_40yrs80occ 0.113      0.098      0.060
## blt40_yrs80occ   0.299      0.238      0.007
## detach80occ      0.858      0.891      0.050
## attach80occ      0.060      0.031      0.041
## mobile80occ      0.081      0.078      0.793
## occupied80       0.941      0.941      0.940
## bltmore30_80     0.409      0.333      0.001
## -----
```

```
site_covariates_2 <- subset(site_covariates, (hrs_82 >= 16.5 & hrs_82 <= 40.5))
group.ttest = function(x, group = as.factor(site_covariates_2$hrs_82 > 28.5)) {
  return(
    unlist(
      t.test(x ~ group)[c("estimate", "p.value")]
    )
  )
}
comp_cov = t(sapply(site_covariates_2[,c(1,3:58)], group.ttest))
colnames(comp_cov) <- c("16.5 to 28.5", "28.5 to 40.5", "p-value")
rownames(comp_cov) <- names(site_covariates_2[c(1,3:58)])
stargazer(comp_cov, title = "Table 1b3: Group by 28.5 for HRS 16.5 - 40.5", type = "text")
```

```
##
## Table 1b3: Group by 28.5 for HRS 16.5 - 40.5
## =====
##              16.5 to 28.5  28.5 to 40.5  p-value
## -----
## hrs_82      22.513      34.890      0
## np12000     0.267      0.985      0
## pop_den8    1,360.901    1,151.050    0.571
## shrblk8     0.082      0.084      0.926
## shrhsp8     0.031      0.030      0.928
## child8      0.288      0.293      0.568
## shrfor8     0.040      0.039      0.785
## ffh8        0.164      0.166      0.862
## smhse8      0.585      0.565      0.244
## hsdrop8     0.148      0.134      0.303
## no_hs_diploma8 0.388      0.353      0.060
## ba_or_better8 0.109      0.134      0.036
## unemprt8    0.075      0.073      0.731
## povrat8     0.107      0.112      0.716
```


## welfare8	0.081	0.075	0.578
## favinc8	21,820.050	22,587.850	0.287
## avhhin8	19,812.210	20,300.790	0.486
## meanrnt80	252.898	256.727	0.643
## mdvalhs9	84,461.760	91,611.450	0.433
## meanrnt9	477.147	481.591	0.842
## mdvalhs0	117,527.800	123,502.900	0.449
## meanrnt0	634.392	621.767	0.677
## tothsun8	1,366.933	1,319.328	0.576
## ownocc8	946.444	871.533	0.244
## owner_occupied80	0.694	0.673	0.344
## bltlast5yrs80	0.137	0.140	0.844
## bltlast10yrs80	0.267	0.276	0.723
## firestoveheat80	0.061	0.055	0.510
## noaircond80	0.516	0.510	0.870
## nofullkitchen80	0.023	0.022	0.787
## zerofullbath80	0.034	0.029	0.386
## northeast	0.389	0.423	0.607
## midwest	0.322	0.285	0.551
## south	0.189	0.204	0.775
## west	0.100	0.088	0.757
## meanhs8	46,135.140	50,647.970	0.084
## bedrms02_80	0.467	0.450	0.434
## bedrms34_80	0.509	0.521	0.557
## detach80	0.693	0.685	0.821
## bedrms0_80occ	0.002	0.003	0.641
## bedrms1_80occ	0.046	0.046	0.897
## bedrms2_80occ	0.308	0.275	0.041
## bedrms3_80occ	0.469	0.480	0.472
## bedrms4_80occ	0.144	0.162	0.089
## bedrms5_80occ	0.030	0.035	0.219
## blt0_1yrs80occ	0.031	0.031	0.960
## blt2_5yrs80occ	0.103	0.102	0.994
## blt6_10yrs80occ	0.120	0.126	0.649
## blt10_20yrs80occ	0.190	0.196	0.687
## blt20_30yrs80occ	0.192	0.193	0.961
## blt30_40yrs80occ	0.108	0.100	0.481
## blt40_yrs80occ	0.255	0.251	0.896
## detach80occ	0.855	0.890	0.108
## attach80occ	0.051	0.032	0.298
## mobile80occ	0.094	0.079	0.285
## occupied80	0.941	0.941	0.989
## bltmore30_80	0.366	0.354	0.662
## -----			

We can see from these above comparisons that with the NPL 2000 treatment vs. control comparison the housing values in 2000 tend to be 8-10% higher in NPL tracts than those with non-NPLs, as expected. Most of the covariates are significantly different between treatment vs. control, which is not ideal. However, the differences are quite small. NPL tracts tend to be less population dense and had families making slightly lower incomes in 1980. The 1982 HRS test result comparisons are less clear, with incomes and housing prices actually looking to be somewhat higher in tracts that had a high 1982 HRS score compared to those that didn't. Therefore, this suggests that perhaps something occurred to cause this change (hence the need for regression discontinuity).

2a - HRS and regression discontinuity

Regression discontinuity design requires that treatment assignment is “as good as random” at the threshold for treatment. If this holds, then it guarantees that those who just barely received treatment are comparable to those who just barely did not receive treatment, as treatment status is effectively random. Treatment assignment at the threshold can be “as good as random” if there is randomness in the assignment variable and the agents considered individuals cannot perfectly manipulate their treatment status.

Regarding the facts:

- i. This fact limits the randomness of the of the assignment variable since the EPA is selecting the number of sites that it can manage to treat. This could suggest that the treated units are not balanced in some way. EPA might have selected the few sites that had the largest potential for benefit. On the other hand, if the threshold was determined for some reason like all HRS scores above 28.5 produce visible or noxious signs of pollution, then there may be other reasons that housing prices might respond to this HRS value other than due to the NPL listing.
- ii. This fact should alleviate concerns that there was gaming around the threshold. The fact that none of the officials involved in site identification, testing and HRS scoring knew about the cutoff threshold at the time means that they were not subject to any bias to score sites near the threshold one way or the other. One can imagine a scenario in which someone may want to score a site with an HRS score of say 29 as being 28.5 so that it might be listed if they have a vested interest in that community being aware of an NPL site. Conversely it is reasonable to think that one might not want to subject a community to an NPL listing if it has an HRS score just below the cutoff so as to prevent housing prices from dropping in response to NPL listing. We could also check this issue by evaluating bunching on our own.
- iii. This fact could potentially muddy the treatment effect on the home’s prices since scores may be off in a way that leads some homes that have really bad pollution not to be treated and homes that don’t have that bad of pollution to be treated (when these homes are close to the threshold). The is a reason for fuzzy RD. The fact that the EPA emphasized that the HRS score is an imperfect scoring measure again means that there may be some slight element of randomness to the assignment and that the HRS score does not reflect detectable characteristics of the tracts other than causing them to be NPL listed or not.

2b - histogram

```
DCdensity(  
  runvar      = mile_data$hrs_82,  
  cutpoint    = 28.5,  
  plot        = TRUE,  
  ext.out     = TRUE)
```

```
## $theta  
## [1] 0.1463121  
##  
## $se  
## [1] 0.3003333  
##  
## $z  
## [1] 0.4871657  
##  
## $p  
## [1] 0.6261409  
##  
## $binsize
```

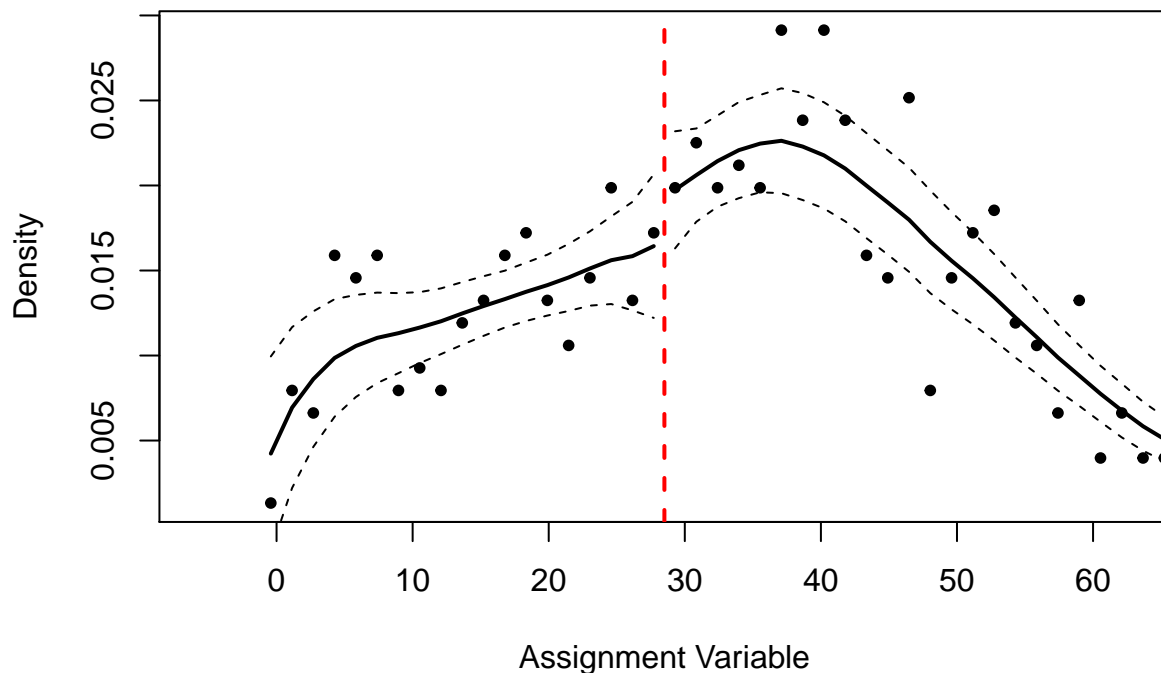
```

## [1] 1.563345
##
## $bw
## [1] 12.21568
##
## $cutpoint
## [1] 28.5
##
## $data
##      cellmp      cellval
## 1  -0.4218786 0.001324336
## 2   1.1414662 0.007946014
## 3   2.7048110 0.006621679
## 4   4.2681558 0.015892029
## 5   5.8315005 0.014567693
## 6   7.3948453 0.015892029
## 7   8.9581901 0.007946014
## 8  10.5215349 0.009270350
## 9  12.0848797 0.007946014
## 10 13.6482245 0.011919022
## 11 15.2115693 0.013243357
## 12 16.7749141 0.015892029
## 13 18.3382589 0.017216365
## 14 19.9016037 0.013243357
## 15 21.4649484 0.010594686
## 16 23.0282932 0.014567693
## 17 24.5916380 0.019865036
## 18 26.1549828 0.013243357
## 19 27.7183276 0.017216365
## 20 29.2816724 0.019865036
## 21 30.8450172 0.022513708
## 22 32.4083620 0.019865036
## 23 33.9717068 0.021189372
## 24 35.5350516 0.019865036
## 25 37.0983963 0.029135386
## 26 38.6617411 0.023838043
## 27 40.2250859 0.029135386
## 28 41.7884307 0.023838043
## 29 43.3517755 0.015892029
## 30 44.9151203 0.014567693
## 31 46.4784651 0.025162379
## 32 48.0418099 0.007946014
## 33 49.6051547 0.014567693
## 34 51.1684995 0.017216365
## 35 52.7318442 0.018540700
## 36 54.2951890 0.011919022
## 37 55.8585338 0.010594686
## 38 57.4218786 0.006621679
## 39 58.9852234 0.013243357
## 40 60.5485682 0.003973007
## 41 62.1119130 0.006621679
## 42 63.6752578 0.003973007
## 43 65.2386026 0.003973007
## 44 66.8019474 0.003973007

```

```
## 45 68.3652921 0.002648671
## 46 69.9286369 0.005297343
## 47 71.4919817 0.003973007
## 48 73.0553265 0.003973007
## 49 74.6186713 0.001324336

title(xlab = 'Assignment Variable', ylab = 'Density')
abline(v = 28.5, col = "red", lty = "dashed", lwd = 2)
```



As we can see in the histogram and in the local linear regression lines, there is a limited discontinuity around the threshold density which suggests we should not be concerned about bunching. Furthermore, the p-value is 0.62 which is further evidence of the insignificance of the discontinuity gap.

3a - 2SLS

First Stage Equation: $d_i = \gamma_0 + \gamma_1 * z_i + \gamma_2 * z_i * (X_i - c) + \gamma_3(X_i - c) + u_i$ Where z_i is the IV, in this case the binary 0/1 variable for whether a tract has an HRS 1982 score above 28.5 (1) or not (0), and d_i is the variable we want to instrument for, in this case the treatment variable of the NPL 2000 listing.

```
mile_data$hrs_82_indicator <- ifelse(mile_data$hrs_82 >= 28.5, 1, 0)
mile_data$running <- mile_data$hrs_82 - 28.5
covariates <- colnames(mile_data)
remove <- c("mdvalhs9_nbr", "mdvalhs9", "mdvalhs0", "meanhs8",
            "lnmdvalhs0_nbr", "fips", "statefips", "npl1990", "hrs_82", "blt40_yrs80occ_nbr")
covariates <- covariates[! covariates %in% remove]
mile_data_reg <- mile_data[, covariates]
```

```

mile_data_reg$interaction <- mile_data_reg$hrs_82_indicator * mile_data_reg$running
reg1 <- lm(npl2000 ~ ., data = mile_data_reg)
summary(reg1)

```

```

##
## Call:
## lm(formula = npl2000 ~ ., data = mile_data_reg)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.96611 -0.07643 -0.00946  0.03896  0.93677
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.754e+05  8.341e+05   0.690  0.49069
## tothsun8_nbr   -1.438e-05  1.645e-05  -0.874  0.38239
## occupied80_nbr  -8.407e-01  4.681e-01  -1.796  0.07321 .
## pop_den8_nbr    -1.832e-07  9.261e-06  -0.020  0.98422
## no_hs_diploma8_nbr -5.041e-01  2.547e-01  -1.979  0.04840 *
## ba_or_better8_nbr -1.893e-01  3.416e-01  -0.554  0.57978
## shrblk8_nbr      1.359e-01  1.523e-01   0.892  0.37294
## shrhsp8_nbr      2.168e-01  2.012e-01   1.077  0.28185
## child8_nbr      -8.106e-01  5.476e-01  -1.480  0.13953
## old8_nbr        -9.230e-01  4.433e-01  -2.082  0.03791 *
## shrfor8_nbr      2.385e-01  3.484e-01   0.684  0.49410
## ffh8_nbr         1.343e-02  2.839e-01   0.047  0.96230
## smhse8_nbr       6.311e-01  2.197e-01   2.872  0.00428 **
## hsdrop8_nbr      -3.799e-02  2.238e-01  -0.170  0.86528
## unemp8_nbr       -1.299e-01  5.275e-01  -0.246  0.80555
## povrat8_nbr       5.852e-01  4.891e-01   1.196  0.23219
## welfare8_nbr     -4.721e-02  5.951e-01  -0.079  0.93681
## avh8_nbr         1.462e-06  5.270e-06   0.277  0.78154
## zerofullbath80_nbr -1.448e+00  1.248e+00  -1.160  0.24661
## firestoveheat80_nbr -2.089e-01  2.400e-01  -0.871  0.38441
## nofullkitchen80_nbr -2.796e-01  1.408e+00  -0.199  0.84264
## noaircond80_nbr    1.510e-01  6.266e-02   2.409  0.01641 *
## ownocc8_nbr       6.743e-06  8.177e-06   0.825  0.41006
## bedrms0_80occ_nbr  7.880e+04  5.716e+05   0.138  0.89041
## bedrms1_80occ_nbr  7.880e+04  5.716e+05   0.138  0.89041
## bedrms2_80occ_nbr  7.880e+04  5.716e+05   0.138  0.89041
## bedrms3_80occ_nbr  7.880e+04  5.716e+05   0.138  0.89041
## bedrms4_80occ_nbr  7.880e+04  5.716e+05   0.138  0.89041
## bedrms5_80occ_nbr  7.880e+04  5.716e+05   0.138  0.89041
## detach80occ_nbr   -6.542e+05  5.619e+05  -1.164  0.24500
## attach80occ_nbr   -6.542e+05  5.619e+05  -1.164  0.24500
## mobile80occ_nbr   -6.542e+05  5.619e+05  -1.164  0.24500
## blt0_1yrs80occ_nbr -3.542e-01  7.018e-01  -0.505  0.61400
## blt2_5yrs80occ_nbr  4.773e-01  3.164e-01   1.509  0.13210
## blt6_10yrs80occ_nbr -1.657e-01  2.620e-01  -0.633  0.52737
## blt10_20yrs80occ_nbr  9.367e-02  1.831e-01   0.512  0.60917
## blt20_30yrs80occ_nbr -1.373e-01  1.660e-01  -0.827  0.40851
## blt30_40yrs80occ_nbr -4.876e-01  2.941e-01  -1.658  0.09802 .
## tothsun9_nbr      2.913e-05  3.203e-05   0.909  0.36366
## tothsun0_nbr     -1.798e-05  2.312e-05  -0.778  0.43718

```

```
## lnmeanhs8_nbr      -4.400e-02  4.915e-02 -0.895  0.37123
## hrs_82_indicator    6.417e-01  4.110e-02 15.612  < 2e-16 ***
## running             1.427e-02  2.304e-03  6.194  1.35e-09 ***
## interaction         -1.455e-02  2.585e-03 -5.631  3.20e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.226 on 439 degrees of freedom
## Multiple R-squared:  0.784, Adjusted R-squared:  0.7628
## F-statistic: 37.05 on 43 and 439 DF,  p-value: < 2.2e-16

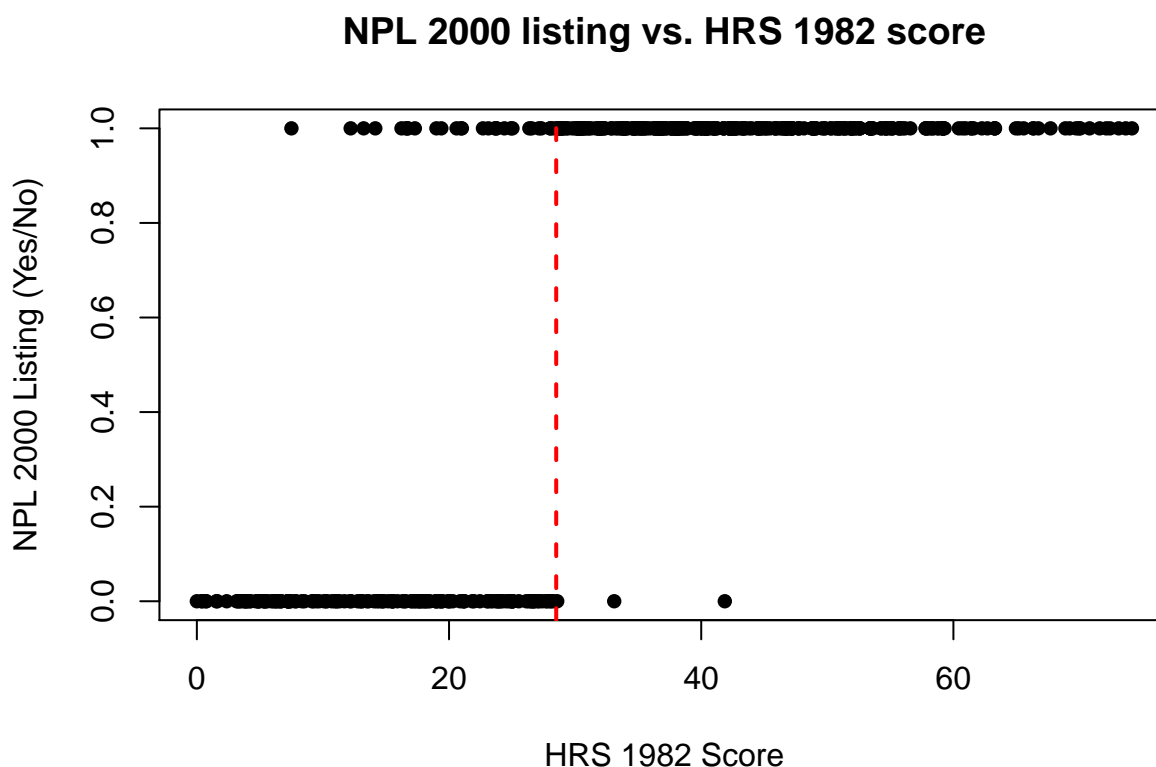
mile_data_reg <- mile_data_reg[which(mile_data_reg$running >= -12 & mile_data_reg$running <= 12),]
reg2_1st <- lm(npl2000 ~ ., data = mile_data_reg)
summary(reg2_1st)

##
## Call:
## lm(formula = npl2000 ~ ., data = mile_data_reg)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.88833 -0.16683 -0.02606  0.07462  0.80964
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.395e+05  1.627e+06   0.209  0.8350
## tothsun8_nbr   -3.463e-05  4.249e-05  -0.815  0.4161
## occupied80_nbr -1.307e+00  1.023e+00  -1.277  0.2031
## pop_den8_nbr    5.627e-06  2.100e-05   0.268  0.7891
## no_hs_diploma8_nbr -4.542e-01  5.194e-01  -0.875  0.3829
## ba_or_better8_nbr -1.901e-01  6.608e-01  -0.288  0.7740
## shrblk8_nbr     -4.753e-02  3.336e-01  -0.142  0.8869
## shrhsp8_nbr     2.895e-01  4.334e-01   0.668  0.5050
## child8_nbr      -1.378e+00  9.513e-01  -1.448  0.1492
## old8_nbr        -1.128e+00  8.317e-01  -1.356  0.1766
## shrfor8_nbr     -1.868e-01  9.667e-01  -0.193  0.8470
## ffh8_nbr        9.577e-01  8.025e-01   1.193  0.2343
## smhse8_nbr      1.124e+00  4.499e-01   2.499  0.0133 *
## hsdrop8_nbr     -3.023e-01  4.795e-01  -0.630  0.5293
## unemp80_nbr     4.959e-01  9.659e-01   0.513  0.6083
## povrat8_nbr     9.404e-01  9.958e-01   0.944  0.3462
## welfare8_nbr    -2.973e-02  1.282e+00  -0.023  0.9815
## avhhin8_nbr     1.160e-05  1.091e-05   1.064  0.2889
## zerofullbath80_nbr -2.835e-01  2.689e+00  -0.105  0.9162
## firestoveheat80_nbr 9.736e-02  5.444e-01   0.179  0.8583
## nofullkitchen80_nbr -3.536e+00  2.945e+00  -1.201  0.2314
## noaircond80_nbr  2.223e-01  1.210e-01   1.837  0.0679 .
## ownocc8_nbr     1.278e-05  1.655e-05   0.772  0.4409
## bedrms0_80occ_nbr -4.094e+05  1.084e+06  -0.378  0.7062
## bedrms1_80occ_nbr -4.094e+05  1.084e+06  -0.378  0.7062
## bedrms2_80occ_nbr -4.094e+05  1.084e+06  -0.378  0.7062
## bedrms3_80occ_nbr -4.094e+05  1.084e+06  -0.378  0.7062
## bedrms4_80occ_nbr -4.094e+05  1.084e+06  -0.378  0.7062
## bedrms5_80occ_nbr -4.094e+05  1.084e+06  -0.378  0.7062
## detach80occ_nbr  6.993e+04  1.137e+06   0.062  0.9510
```

```
## attach80occ_nbr      6.993e+04  1.137e+06   0.062   0.9510
## mobile80occ_nbr      6.993e+04  1.137e+06   0.062   0.9510
## blt0_1yrs80occ_nbr  -1.148e+00  1.560e+00  -0.736   0.4627
## blt2_5yrs80occ_nbr   1.547e+00  7.266e-01   2.130   0.0345 *
## blt6_10yrs80occ_nbr  2.532e-01  5.691e-01   0.445   0.6569
## blt10_20yrs80occ_nbr 3.191e-01  4.200e-01   0.760   0.4484
## blt20_30yrs80occ_nbr -2.631e-01  3.360e-01  -0.783   0.4347
## blt30_40yrs80occ_nbr -3.060e-01  6.279e-01  -0.487   0.6267
## tothsun9_nbr         6.202e-05  7.720e-05   0.803   0.4228
## tothsun0_nbr        -3.348e-05  5.122e-05  -0.654   0.5142
## lnmeanhs8_nbr       -4.095e-02  9.000e-02  -0.455   0.6496
## hrs_82_indicator     6.287e-01  8.549e-02   7.354 6.33e-12 ***
## running              1.705e-02  9.321e-03   1.830   0.0690 .
## interaction          -1.702e-02  1.191e-02  -1.429   0.1547
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2955 on 182 degrees of freedom
## Multiple R-squared:  0.6629, Adjusted R-squared:  0.5832
## F-statistic: 8.322 on 43 and 182 DF,  p-value: < 2.2e-16
```

3b - graph results

```
plot(mile_data$hrs_82, mile_data$np12000, main = "NPL 2000 listing vs. HRS 1982 score",
     pch = 16, ylab = "NPL 2000 Listing (Yes/No)", xlab = "HRS 1982 Score")
abline(v = 28.5, lty = 'dashed', col = 2, lwd = 2)
```

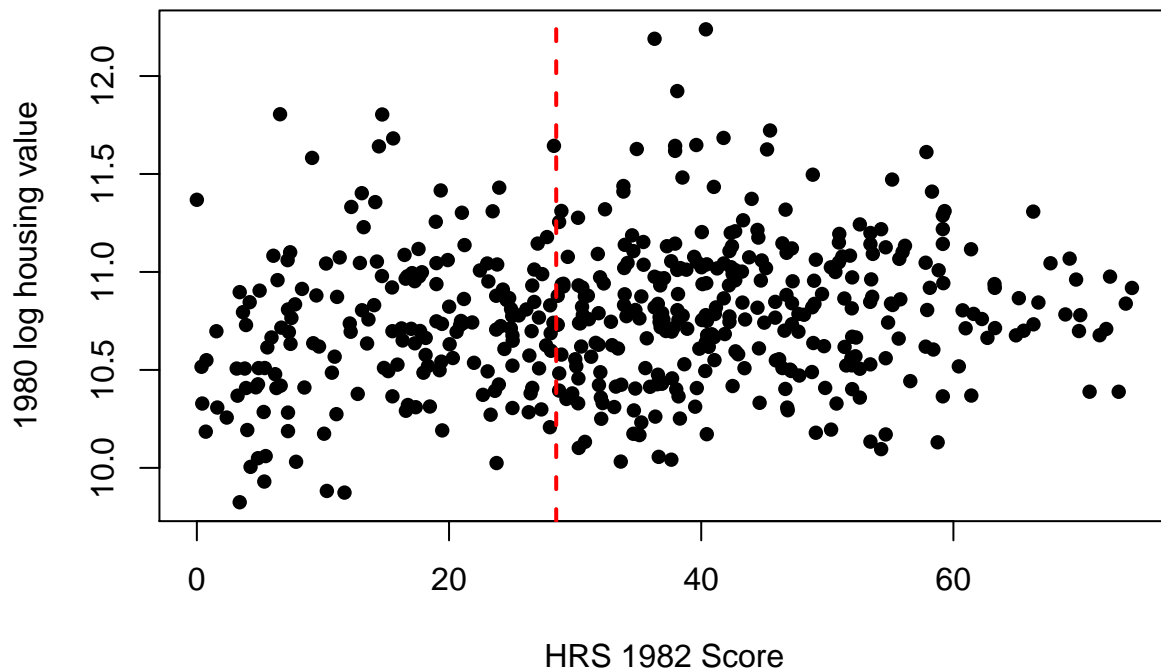


The graph shows a clear trend of houses with scores less than 28.5 in 1982 being classified as non-priority (NPL value of 0) in 2000, while houses with scores greater than 28.5 in 1982 are on the National Priorities List (NPL value of 1) for clean-up. There are also a few houses with scores lower than 28.5 that are on the NPL in 2000 showing that there is an increase in the number of hazardous waste sites that are legally obligated to be cleaned up since 1982, when their initial scores were assigned under the Superfund program. This shows that there is a fuzzy discontinuity in NPL classification at an HRS score of 28.5 which can be used in a regression discontinuity design study to study the differential effects of hazardous waste material above a certain threshold on housing prices. The fact that there is a correlation suggests that the first assumption needed to use HRS 82 scores as an IV is valid because clearly the covariance between HRS 82 scores and NPL 2000 listing is not equal to zero, or $Cov(z_i, d_i) \neq 0$.

3c - graph results #2

```
plot(mile_data$hrs_82, mile_data$lnmeanhs8_nbr, main = "1980 housing value vs. HRS 1982 score",
     pch = 16, ylab = "1980 log housing value", xlab = "HRS 1982 Score")
abline(v = 28.5, lty = 'dashed', col = 2, lwd = 2)
```


1980 housing value vs. HRS 1982 score



There is no clear trend between the HRS score in 1982 and mean housing prices in 1980. This shows that ostensibly there may not be a relationship between the two variables prior to the Superfund program, which is desirable because we would not want a covariate to have a regression discontinuity.

4 - More 2SLS

reduced form Equation: $y_i = \pi_0 + \pi_1 z_i + \pi_2(x_i - c)z_i + \pi_2(x_i - c) + \varepsilon_i$ Where y_i is the outcome variable, in this case the difference in log housing value between 1980 and 2000, and z_i is the instrument, and x_i are the rest of the exogenous covariates. In the second stage, we regress the outcome variable on predicted values from the first stage (\hat{x}_{star}) and all the exogenous variables in the model. The 2 standard assumptions for a valid IV are as follows: 1. It is uncorrelated with the disturbance - $Cov(\text{instrument}, \text{error}) = 0$ 2. In the linear projection of the endogenous variable on all exogenous variables and the instrument, the coefficient on the instrument is not equal to 0. This implies that the only way the instrument affects the outcome variable is through the endogenous variable. The third condition is a bit redundant, but we will state it anyway. The instrument is not one of the other variables in the original estimation equation.

```
covariates <- colnames(mile_data)
remove <- c("mdvalhs9_nbr", "mdvalhs9", "mdvalhs0", "meanhs8",
            "fips", "statefips", "npl1990", "hrs_82", "npl2000", "blt40_yrs80occ_nbr")
covariates <- covariates[! covariates %in% remove]
mile_data_2sls <- mile_data[,covariates]
mile_data_2sls$interaction <- mile_data_2sls$hrs_82_indicator * mile_data_2sls$running
reg1 <- lm(lnmdvalhs0_nbr ~ ., data = mile_data_2sls)
summary(reg1)
```

```
##
```

```
## Call:
## lm(formula = lnmdvalhs0_nbr ~ ., data = mile_data_2sls)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.71261 -0.11227  0.01351  0.11088  0.75039
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -8.793e+05  7.337e+05  -1.199  0.231347
## tothsun8_nbr    3.489e-05  1.446e-05   2.412  0.016285 *
## occupied80_nbr   1.419e+00  4.117e-01   3.447  0.000621 ***
## pop_den8_nbr    -2.024e-06  8.146e-06  -0.248  0.803882
## no_hs_diploma8_nbr -1.611e-01  2.240e-01  -0.719  0.472538
## ba_or_better8_nbr  3.391e-01  3.005e-01   1.128  0.259743
## shrblk8_nbr     -3.247e-01  1.340e-01  -2.423  0.015803 *
## shrhsp8_nbr     -3.875e-01  1.770e-01  -2.190  0.029071 *
## child8_nbr      -1.077e+00  4.817e-01  -2.236  0.025879 *
## old8_nbr        -8.285e-01  3.899e-01  -2.125  0.034159 *
## shrfor8_nbr      6.906e-01  3.065e-01   2.253  0.024730 *
## ffh8_nbr        6.379e-01  2.497e-01   2.555  0.010965 *
## smhse8_nbr      5.400e-01  1.933e-01   2.794  0.005433 **
## hsdrop8_nbr      4.968e-01  1.968e-01   2.524  0.011954 *
## unemprt8_nbr    -1.962e+00  4.640e-01  -4.230  2.85e-05 ***
## povrat8_nbr     -1.137e-02  4.302e-01  -0.026  0.978917
## welfare8_nbr     2.074e+00  5.234e-01   3.962  8.67e-05 ***
## avhhin8_nbr      4.220e-05  4.635e-06   9.106  < 2e-16 ***
## zerofullbath80_nbr 3.911e-01  1.098e+00   0.356  0.721871
## firestoveheat80_nbr 4.655e-01  2.111e-01   2.205  0.027957 *
## nofullkitchen80_nbr -1.104e+00  1.238e+00  -0.892  0.373033
## noaircond80_nbr   2.249e-01  5.512e-02   4.081  5.32e-05 ***
## ownocc8_nbr     -2.552e-05  7.192e-06  -3.548  0.000430 ***
## bedrms0_80occ_nbr 7.098e+05  5.027e+05   1.412  0.158690
## bedrms1_80occ_nbr 7.098e+05  5.027e+05   1.412  0.158685
## bedrms2_80occ_nbr 7.098e+05  5.027e+05   1.412  0.158685
## bedrms3_80occ_nbr 7.098e+05  5.027e+05   1.412  0.158685
## bedrms4_80occ_nbr 7.098e+05  5.027e+05   1.412  0.158686
## bedrms5_80occ_nbr 7.098e+05  5.027e+05   1.412  0.158685
## detach80occ_nbr   1.695e+05  4.943e+05   0.343  0.731742
## attach80occ_nbr   1.695e+05  4.943e+05   0.343  0.731743
## mobile80occ_nbr   1.695e+05  4.943e+05   0.343  0.731742
## blt0_1yrs80occ_nbr 2.536e-01  6.173e-01   0.411  0.681438
## blt2_5yrs80occ_nbr 2.957e-02  2.783e-01   0.106  0.915414
## blt6_10yrs80occ_nbr 3.502e-01  2.305e-01   1.520  0.129321
## blt10_20yrs80occ_nbr -2.514e-01  1.610e-01  -1.561  0.119231
## blt20_30yrs80occ_nbr -8.850e-02  1.460e-01  -0.606  0.544665
## blt30_40yrs80occ_nbr -2.597e-01  2.587e-01  -1.004  0.315906
## tothsun9_nbr     -1.087e-04  2.817e-05  -3.858  0.000132 ***
## tothsun0_nbr      8.701e-05  2.033e-05   4.280  2.30e-05 ***
## lnmeanhs8_nbr     5.299e-01  4.323e-02  12.257  < 2e-16 ***
## hrs_82_indicator  1.889e-03  3.615e-02   0.052  0.958346
## running          -3.082e-04  2.027e-03  -0.152  0.879211
## interaction       8.251e-04  2.273e-03   0.363  0.716847
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1988 on 439 degrees of freedom
## Multiple R-squared:  0.8156, Adjusted R-squared:  0.7975
## F-statistic: 45.15 on 43 and 439 DF,  p-value: < 2.2e-16

mile_data_2sls <- mile_data_2sls[which(mile_data_2sls$running >= -12 & mile_data_2sls$running <= 12),]
reg2 <- lm(lnmdvalhs0_nbr ~ ., data = mile_data_2sls)
summary(reg2)

##
## Call:
## lm(formula = lnmdvalhs0_nbr ~ ., data = mile_data_2sls)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.42902 -0.12521 -0.00043  0.12517  0.49872
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -6.407e+05  1.144e+06  -0.560  0.57618
## tothsun8_nbr    -1.958e-06  2.987e-05  -0.066  0.94782
## occupied80_nbr     2.242e+00  7.193e-01   3.117  0.00212 **
## pop_den8_nbr     -2.904e-06  1.477e-05  -0.197  0.84429
## no_hs_diploma8_nbr -3.544e-01  3.651e-01  -0.971  0.33297
## ba_or_better8_nbr -3.221e-01  4.645e-01  -0.693  0.48896
## shrblk8_nbr      -1.665e-01  2.345e-01  -0.710  0.47858
## shrhsp8_nbr       3.148e-01  3.047e-01   1.033  0.30286
## child8_nbr       -1.305e+00  6.688e-01  -1.951  0.05263 .
## old8_nbr         -9.120e-01  5.847e-01  -1.560  0.12056
## shrfor8_nbr       1.145e+00  6.796e-01   1.685  0.09379 .
## ffh8_nbr         5.876e-03  5.641e-01   0.010  0.99170
## smhse8_nbr       3.145e-01  3.163e-01   0.994  0.32145
## hsdrop8_nbr       1.499e-01  3.371e-01   0.445  0.65712
## unemprt8_nbr     -1.766e+00  6.790e-01  -2.601  0.01007 *
## povrat8_nbr       6.695e-01  7.000e-01   0.956  0.34017
## welfare8_nbr      1.604e+00  9.013e-01   1.780  0.07679 .
## avhhin8_nbr       4.610e-05  7.667e-06   6.012 9.81e-09 ***
## zerofullbath80_nbr -6.193e-01  1.891e+00  -0.328  0.74359
## firestoveheat80_nbr  8.181e-01  3.827e-01   2.138  0.03389 *
## nofullkitchen80_nbr  3.204e-01  2.071e+00   0.155  0.87721
## noaircond80_nbr    2.116e-01  8.509e-02   2.487  0.01379 *
## ownocc8_nbr      -5.347e-06  1.163e-05  -0.460  0.64632
## bedrms0_80occ_nbr  7.874e+05  7.623e+05   1.033  0.30302
## bedrms1_80occ_nbr  7.874e+05  7.623e+05   1.033  0.30301
## bedrms2_80occ_nbr  7.874e+05  7.623e+05   1.033  0.30301
## bedrms3_80occ_nbr  7.874e+05  7.623e+05   1.033  0.30301
## bedrms4_80occ_nbr  7.874e+05  7.623e+05   1.033  0.30301
## bedrms5_80occ_nbr  7.874e+05  7.623e+05   1.033  0.30301
## detach80occ_nbr   -1.467e+05  7.993e+05  -0.184  0.85458
## attach80occ_nbr   -1.467e+05  7.993e+05  -0.184  0.85458
## mobile80occ_nbr   -1.467e+05  7.993e+05  -0.184  0.85458
## blt0_1yrs80occ_nbr -1.016e-01  1.097e+00  -0.093  0.92631
## blt2_5yrs80occ_nbr  1.360e-01  5.108e-01   0.266  0.79029
## blt6_10yrs80occ_nbr 1.671e-01  4.001e-01   0.418  0.67666
```

```
## blt10_20yrs80occ_nbr -3.040e-01 2.953e-01 -1.030 0.30458
## blt20_30yrs80occ_nbr -1.713e-01 2.362e-01 -0.725 0.46914
## blt30_40yrs80occ_nbr -8.920e-01 4.414e-01 -2.021 0.04478 *
## tothsun9_nbr -5.527e-05 5.427e-05 -1.018 0.30981
## tothsun0_nbr 6.258e-05 3.601e-05 1.738 0.08392 .
## lnmeanhs8_nbr 3.812e-01 6.327e-02 6.025 9.18e-09 ***
## hrs_82_indicator 5.924e-02 6.010e-02 0.986 0.32562
## running -1.886e-03 6.553e-03 -0.288 0.77387
## interaction -6.125e-03 8.375e-03 -0.731 0.46554
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2077 on 182 degrees of freedom
## Multiple R-squared: 0.8051, Adjusted R-squared: 0.759
## F-statistic: 17.48 on 43 and 182 DF, p-value: < 2.2e-16
```

```
##calculate treatment effect
```

```
reg2$coefficients[[42]] / reg2_1st$coefficients[[42]]
```

```
## [1] 0.09422333
```

```
##calculate treatment effect with 1 regression
```

```
mile_data_2sls$fitted <- reg2_1st$fitted.values
```

```
covariates <- colnames(mile_data_2sls)
```

```
remove <- c("mdvalhs9_nbr", "mdvalhs9", "mdvalhs0", "meanhs8",
            "fips", "statefips", "npl1990", "hrs_82", "npl2000", "blt40_yrs80occ_nbr",
            "hrs_82_indicator")
```

```
covariates_2sls <- covariates[! covariates %in% remove]
```

```
mile_data_2sls <- mile_data_2sls[,covariates_2sls]
```

```
mile_data_2sls$interaction <- mile_data_2sls$fitted * mile_data_2sls$running
```

```
reg_2sls <- lm(lnmdvalhs0_nbr ~ ., data = mile_data_2sls)
```

```
summary(reg_2sls)
```

```
##
```

```
## Call:
```

```
## lm(formula = lnmdvalhs0_nbr ~ ., data = mile_data_2sls)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -0.43290 -0.12584  0.00115  0.12550  0.50203
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -6.403e+05  1.143e+06  -0.560  0.57601
## tothsun8_nbr    1.896e-06  2.996e-05   0.063  0.94960
## occupied80_nbr   2.377e+00  7.297e-01   3.257  0.00134 **
## pop_den8_nbr    -3.669e-06  1.484e-05  -0.247  0.80509
## no_hs_diploma8_nbr -3.102e-01  3.681e-01  -0.843  0.40044
## ba_or_better8_nbr -3.013e-01  4.683e-01  -0.643  0.52076
## shrblk8_nbr     -1.573e-01  2.352e-01  -0.669  0.50433
## shrhsp8_nbr      2.868e-01  3.061e-01   0.937  0.35014
## child8_nbr      -1.162e+00  6.996e-01  -1.660  0.09855 .
## old8_nbr        -7.796e-01  5.995e-01  -1.300  0.19512
## shrfor8_nbr      1.158e+00  6.801e-01   1.702  0.09038 .
## ffh8_nbr        -1.039e-01  5.774e-01  -0.180  0.85741
## smhse8_nbr       1.822e-01  3.249e-01   0.561  0.57570
```

```

## hsdrop8_nbr          1.824e-01  3.409e-01  0.535  0.59334
## unemprt8_nbr         -1.841e+00  6.805e-01 -2.705  0.00748 **
## povrat8_nbr          5.459e-01  7.131e-01  0.766  0.44491
## welfare8_nbr         1.610e+00  9.027e-01  1.783  0.07624 .
## avhhin8_nbr          4.480e-05  7.882e-06  5.684  5.16e-08 ***
## zerofullbath80_nbr   -5.486e-01  1.898e+00 -0.289  0.77284
## firestoveheat80_nbr  7.987e-01  3.840e-01  2.080  0.03891 *
## nofullkitchen80_nbr  7.186e-01  2.114e+00  0.340  0.73430
## noaircond80_nbr      1.874e-01  8.673e-02  2.160  0.03205 *
## ownocc8_nbr          -7.144e-06  1.162e-05 -0.615  0.53954
## bedrms0_80occ_nbr    8.217e+05  7.658e+05  1.073  0.28472
## bedrms1_80occ_nbr    8.217e+05  7.658e+05  1.073  0.28471
## bedrms2_80occ_nbr    8.217e+05  7.658e+05  1.073  0.28471
## bedrms3_80occ_nbr    8.217e+05  7.658e+05  1.073  0.28471
## bedrms4_80occ_nbr    8.217e+05  7.658e+05  1.073  0.28471
## bedrms5_80occ_nbr    8.217e+05  7.658e+05  1.073  0.28471
## detach80occ_nbr      -1.814e+05  8.016e+05 -0.226  0.82118
## attach80occ_nbr      -1.814e+05  8.016e+05 -0.226  0.82118
## mobile80occ_nbr      -1.814e+05  8.016e+05 -0.226  0.82119
## blt0_1yrs80occ_nbr   -4.033e-03  1.108e+00 -0.004  0.99710
## blt2_5yrs80occ_nbr   -3.396e-02  5.307e-01 -0.064  0.94905
## blt6_10yrs80occ_nbr  1.320e-01  4.048e-01  0.326  0.74482
## blt10_20yrs80occ_nbr -3.352e-01  2.984e-01 -1.124  0.26269
## blt20_30yrs80occ_nbr -1.476e-01  2.375e-01 -0.622  0.53499
## blt30_40yrs80occ_nbr -8.513e-01  4.419e-01 -1.926  0.05560 .
## tothsun9_nbr         -6.311e-05  5.451e-05 -1.158  0.24848
## tothsun0_nbr         6.738e-05  3.615e-05  1.864  0.06394 .
## lnmeanhs8_nbr        3.826e-01  6.354e-02  6.021  9.36e-09 ***
## running              -5.467e-03  8.819e-03 -0.620  0.53610
## interaction          -1.591e-03  9.652e-03 -0.165  0.86926
## fitted               1.068e-01  9.374e-02  1.139  0.25613
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2079 on 182 degrees of freedom
## Multiple R-squared:  0.8048, Adjusted R-squared:  0.7587
## F-statistic: 17.45 on 43 and 182 DF,  p-value: < 2.2e-16

```

When running the two regressions separately, we calculate a treatment effect of 0.09. While the number is slightly positive, the results are insignificant. This implies that housing prices in 2000 are insignificantly affected by the NPL 2000 status. When running the 2sls regression completely, the point estimate was very close, at 0.1, but also insignificant.

5 - Summary

We find that a simple OLS regression of housing prices on NPL 2000 status reveals a significant positive relationship when none of the control variables are included in the model. The relationship becomes less significant on adding the housing and demographic characteristics to the equation. We try to resolve omitted variable bias using an IV model as suggested, using the HRS score in 1982 as an instrument for NPL 2000 status. This reveals an insignificant relationship between pollution levels indicated by inclusion on the NPL 2000 list and housing prices. The goal of the paper is to test the Market Willingness to Pay (WTP) for hazardous waste site cleanup using housing prices, and the results of the IV approach reveal that home-owners do not yet value the property on the basis of pollution levels or hazardous waste cleanup status.