### Project 4

### Carra Wu

```
#Change this to location of your data
#Can use drop down menu in R studio: file->import data set-> from stata and find stata data set
setwd(dir = "/Users/carrawu/Documents/harvard/ec1152")
if (!require(foreign)) install.packages("foreign"); library(foreign)
## Loading required package: foreign
if (!require(haven)) install.packages("haven"); library(haven)
## Loading required package: haven
if (!require(randomForest)) install.packages("randomForest"); library(randomForest)
## Loading required package: randomForest
## randomForest 4.6-14
## Type rfNews() to see new features/changes/bug fixes.
if (!require(rpart)) install.packages("rpart"); library(rpart)
## Loading required package: rpart
#Open stata data set
proj4 <- read_dta("project4.dta")</pre>
head(proj4)
## # A tibble: 6 x 144
                   pop housing kfr_pooled_p25 test training rank_hat_ols
##
     geoid place
     <dbl> <chr> <dbl>
                          <dbl>
                                         <dbl> <dbl>
                                                        <dbl>
                                                                     <dbl>
## 1 1001 Auta~ 54907
                          22135
                                        NA
                                                            0
                                                                     0.399
## 2 1003 Bald~ 187114
                         104061
                                         0.389
                                                                     0.395
                                                            1
## 3 1005 Barb~ 27321
                          11829
                                         0.349
                                                   0
                                                            1
                                                                     0.383
## 4 1007 Bibb~ 22754
                                         0.363
                                                                     0.416
                           8981
                                                   0
                                                            1
## 5 1009 Blou~ 57623
                          23887
                                                   1
                                                            0
                                                                     0.392
                                        NA
## 6 1011 Bull~ 10746
                           4493
                                        NA
                                                   1
                                                            0
                                                                     0.427
```

```
## # ... with 136 more variables: personageyears5onwardslang <dbl>, v4 <dbl>,
## #
       personageyears25onwardsedu <dbl>, v6 <dbl>,
## #
       personageyears18onwardsarm <dbl>, v8 <dbl>,
       housingunitoccupancystatus <dbl>, housingunitcashrentstatusw <dbl>,
## #
## #
       personageyears5onwardsnati <dbl>, persongenderfemale <dbl>, v13 <dbl>,
       v14 <dbl>, v15 <dbl>, v16 <dbl>, P_1 <dbl>, P_2 <dbl>, P_3 <dbl>,
## #
## #
       P_4 <dbl>, P_5 <dbl>, P_6 <dbl>, P_7 <dbl>, P_8 <dbl>, P_9 <dbl>,
## #
       P_10 <dbl>, P_11 <dbl>, P_12 <dbl>, P_13 <dbl>, P_14 <dbl>,
## #
       P_15 <dbl>, P_16 <dbl>, P_17 <dbl>, P_18 <dbl>, P_19 <dbl>,
## #
       P_20 <dbl>, P_21 <dbl>, P_22 <dbl>, P_23 <dbl>, P_24 <dbl>,
       P_25 <dbl>, P_26 <dbl>, P_27 <dbl>, P_28 <dbl>, P_29 <dbl>,
## #
## #
       P_30 <dbl>, P_31 <dbl>, P_32 <dbl>, P_33 <dbl>, P_34 <dbl>,
## #
       P_35 <dbl>, P_36 <dbl>, P_37 <dbl>, P_38 <dbl>, P_39 <dbl>,
## #
       P_40 <dbl>, P_41 <dbl>, P_42 <dbl>, P_43 <dbl>, P_44 <dbl>,
       P_45 <dbl>, P_46 <dbl>, P_47 <dbl>, P_48 <dbl>, P_49 <dbl>,
## #
## #
       P_50 <dbl>, P_51 <dbl>, P_52 <dbl>, P_53 <dbl>, P_54 <dbl>,
       P_55 <dbl>, P_56 <dbl>, P_57 <dbl>, P_58 <dbl>, P_59 <dbl>,
## #
       P_60 <dbl>, P_61 <dbl>, P_62 <dbl>, P_63 <dbl>, P_64 <dbl>,
## #
## #
       P_65 <dbl>, P_66 <dbl>, P_67 <dbl>, P_68 <dbl>, P_69 <dbl>,
## #
       P_70 <dbl>, P_71 <dbl>, P_72 <dbl>, P_73 <dbl>, P_74 <dbl>,
## #
       P_75 <dbl>, P_76 <dbl>, P_77 <dbl>, P_78 <dbl>, P_79 <dbl>,
## #
       P_80 <dbl>, P_81 <dbl>, P_82 <dbl>, P_83 <dbl>, P_84 <dbl>,
       P_85 <dbl>, P_86 <dbl>, ...
#Storing predictor variables
#Order data in stata so all predictors appear in right-most columns
vars <- colnames(proj4[10:ncol(proj4)])</pre>
#OLS Regression
to_hat <- with(proj4[proj4$training==1,], lm(reformulate(vars, "kfr_pooled_p25")))</pre>
summary(to_hat)
```

2

##

## Call:

```
## lm(formula = reformulate(vars, "kfr_pooled_p25"))
##
## Residuals:
        Min
                   1Q
                         Median
                                       3Q
##
                                                Max
## -0.101989 -0.010533 -0.000746 0.010740 0.116464
##
## Coefficients: (1 not defined because of singularities)
##
                               Estimate Std. Error t value Pr(>|t|)
                             -4.032e+00 1.025e+01 -0.393 0.694214
## (Intercept)
## v4
                             -1.119e-01 5.880e-02 -1.902 0.057383 .
## personageyears25onwardsedu -5.380e-01 2.294e-01 -2.345 0.019181 *
                                                     4.190 3.01e-05 ***
## v6
                              1.249e-01 2.981e-02
## personageyears18onwardsarm -1.850e-01 4.692e-02 -3.943 8.53e-05 ***
## v8
                             -6.327e-02 3.999e-02 -1.582 0.113929
## housingunitoccupancystatus 2.254e-02 1.200e-02 1.878 0.060674 .
## housingunitcashrentstatusw 1.718e-04 1.681e-02
                                                     0.010 0.991848
## personageyears5onwardsnati 5.225e-02 6.501e-02
                                                     0.804 0.421719
                              2.068e-02 3.772e-02
                                                     0.548 0.583615
## persongenderfemale
## v13
                             -2.875e-02 1.419e-01 -0.203 0.839472
## v14
                             -1.239e+02 3.884e+01 -3.189 0.001465 **
                             -4.914e-01 9.405e-02 -5.225 2.07e-07 ***
## v15
                              1.283e-01 5.231e-02
                                                     2.452 0.014365 *
## v16
## P_1
                             -1.707e-04 8.234e-05 -2.073 0.038366 *
## P_2
                              1.248e-05 9.140e-05
                                                    0.137 0.891423
## P 3
                              1.151e-04 9.379e-05
                                                     1.227 0.220178
## P_4
                              5.625e-05 1.043e-04
                                                     0.539 0.589846
## P 5
                             -1.686e-04 1.124e-04 -1.499 0.134067
## P_6
                             -3.209e-05 1.126e-04 -0.285 0.775689
## P_7
                              1.288e-05 7.286e-05
                                                     0.177 0.859712
## P_8
                             -7.070e-05 1.262e-04 -0.560 0.575318
## P_9
                              1.520e-04 1.111e-04 1.368 0.171433
## P_10
                             -5.101e-03 2.851e-03 -1.789 0.073824 .
## P_11
                             -1.673e-03 3.739e-03 -0.448 0.654563
```

```
## P_12
                            -9.278e-04 5.465e-04 -1.698 0.089851 .
## P_13
                             1.068e-03 5.671e-04 1.883 0.059945 .
## P_14
                             9.469e-04 5.475e-04 1.729 0.084009 .
## P_15
                             9.178e-04 5.481e-04 1.675 0.094289 .
## P_16
                             9.290e-04 5.464e-04
                                                   1.700 0.089391 .
## P_17
                             8.405e-04 8.880e-04
                                                   0.947 0.344079
## P_18
                             8.212e-05 6.079e-04 0.135 0.892565
## P_19
                             8.358e-05 6.078e-04 0.138 0.890642
## P_20
                             1.260e-04 6.076e-04 0.207 0.835722
## P_21
                             2.088e-04 4.709e-04 0.443 0.657584
## P_22
                            -1.998e-04 4.709e-04 -0.424 0.671429
## P_23
                            -2.090e-04 4.708e-04 -0.444 0.657094
## P_24
                            -1.272e-06 2.376e-07 -5.354 1.04e-07 ***
## P_25
                             3.737e-05 2.603e-04 0.144 0.885894
## P_26
                             2.021e-01 4.154e-02 4.866 1.30e-06 ***
## P_27
                            -4.584e-02 4.134e-02 -1.109 0.267728
                            -3.699e-02 6.693e-02 -0.553 0.580628
## P_28
                             1.349e-06 2.134e-07 6.322 3.72e-10 ***
## P_29
## P_30
                            -1.604e-06 4.644e-07 -3.454 0.000572 ***
## P_31
                            -5.875e-02 2.797e-02 -2.101 0.035904 *
                             1.488e-01 4.221e-02 3.526 0.000439 ***
## P_32
## P_33
                            -4.395e-02 3.586e-02 -1.226 0.220636
## P_34
                            -8.358e-02 1.583e-01 -0.528 0.597528
## P_35
                            -1.819e-01 1.541e-01 -1.180 0.238203
## P_36
                            -1.245e-01 2.354e-01 -0.529 0.596908
## P_37
                             1.436e-01 1.694e-01 0.848 0.396673
## P_38
                             2.611e-01 1.665e-01 1.568 0.117152
## P_39
                             2.321e-01 1.653e-01 1.404 0.160728
## P_40
                             3.883e-01 2.642e-01 1.470 0.141955
                             9.222e-04 1.187e-03 0.777 0.437333
## P_41
## P_42
                             6.068e-06 8.782e-06
                                                    0.691 0.489712
## P_43
                             -3.600e-02 1.537e-02 -2.342 0.019340 *
## P_44
                            -3.674e-02 3.033e-02 -1.211 0.226121
```

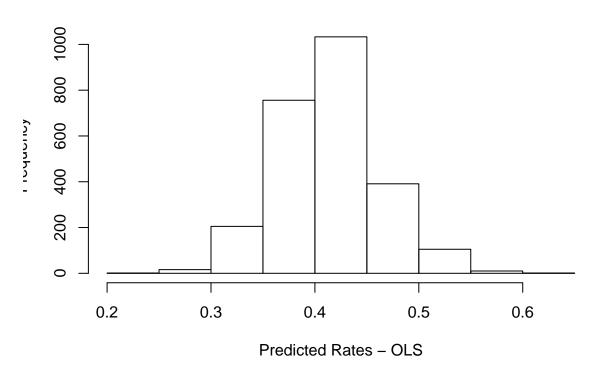
##	P_45	-1.623e-01	3.655e-02	-4.441	9.83e-06	***
##	P_46	4.888e-02	1.171e-02	4.175	3.20e-05	***
##	P_47	-2.778e-02	2.170e-02	-1.280	0.200810	
##	P_48	1.820e-04	1.514e-04	1.202	0.229671	
##	P_49	-1.287e-02	6.108e-03	-2.108	0.035273	*
##	P_50	1.850e-01	1.557e-01	1.188	0.234992	
##	P_51	7.369e-06	2.566e-06	2.872	0.004160	**
##	P_52	-5.980e-06	2.423e-06	-2.468	0.013737	*
##	P_53	-1.481e-01	6.425e-02	-2.305	0.021369	*
##	P_54	-2.309e-06	4.279e-06	-0.540	0.589609	
##	P_55	-7.360e-03	4.733e-03	-1.555	0.120223	
##	P_56	-1.274e-02	3.599e-03	-3.539	0.000418	***
##	P_57	2.162e-03	8.067e-04	2.680	0.007470	**
##	P_58	-9.590e-04	6.794e-04	-1.411	0.158383	
##	P_59	2.406e-05	3.156e-05	0.762	0.445985	
##	P_60	6.726e-06	6.072e-06	1.108	0.268202	
##	P_61	1.821e-05	3.700e-05	0.492	0.622737	
##	P_62	-2.743e-07	6.868e-07	-0.399	0.689721	
##	P_63	3.684e-03	9.602e-04	3.837	0.000131	***
##	P_64	5.507e-03	2.113e-03	2.607	0.009264	**
##	P_65	-5.894e-03	3.897e-03	-1.513	0.130683	
##	P_66	-1.741e-05	1.704e-05	-1.022	0.307101	
##	P_67	-8.380e-05	1.096e-04	-0.764	0.444831	
##	P_68	2.764e-03	1.016e-03	2.719	0.006644	**
##	P_69	-2.900e-04	1.513e-04	-1.916	0.055566	
##	P_70	1.716e-07	2.394e-07	0.717	0.473810	
##	P_71	-3.238e-04	8.392e-05	-3.859	0.000121	***
##	P_72	1.468e-05	5.851e-06	2.510	0.012224	*
##	P_73	1.423e-06	4.099e-06	0.347	0.728513	
##	P_74	-8.738e-04	4.846e-04	-1.803	0.071657	
##	P_75	3.237e-04	3.918e-04	0.826	0.408933	
##	P_76	7.576e-04	2.644e-04	2.865	0.004245	**
##	P_77	-6.525e-05	4.582e-05	-1.424	0.154772	

"" P 70	0.044.04	0 500 04	4 045	0 000500	
## P_78	-2.641e-04				
## P_79	-1.684e-04				
## P_80	-1.192e-04				
## P_81	5.329e-03	2.053e-03	2.596	0.009559	**
## P_82	7.086e-02	6.705e-02	1.057	0.290858	
## P_83	7.102e-02	6.705e-02	1.059	0.289716	
## P_84	6.942e-02	6.707e-02	1.035	0.300938	
## P_85	7.228e-02	6.705e-02	1.078	0.281274	
## P_86	7.434e-02	6.708e-02	1.108	0.268003	
## P_87	7.095e-02	6.706e-02	1.058	0.290263	
## P_88	6.526e-02	6.717e-02	0.972	0.331463	
## P_89	6.443e-02	6.727e-02	0.958	0.338371	
## P_90	7.434e-02	6.722e-02	1.106	0.268957	
## P_91	7.183e-02	6.716e-02	1.070	0.285066	
## P_92	7.091e-02	6.703e-02	1.058	0.290296	
## P_93	7.078e-02	6.705e-02	1.056	0.291423	
## P_94	-2.800e-02	8.006e-02	-0.350	0.726616	
## P_95	-2.803e-02	8.007e-02	-0.350	0.726357	
## P_96	-2.727e-02	8.006e-02	-0.341	0.733508	
## P_97	-2.804e-02	8.007e-02	-0.350	0.726268	
## P_98	-3.043e-02	8.012e-02	-0.380	0.704199	
## P_99	-2.772e-02	8.006e-02	-0.346	0.729279	
## P_100	-2.278e-02	8.026e-02	-0.284	0.776635	
## P_101	-2.447e-02	8.005e-02	-0.306	0.759958	
## P_102	-3.694e-02	8.010e-02	-0.461	0.644744	
## P_103	-2.923e-02	8.006e-02	-0.365	0.715071	
## P_104	-2.812e-02	8.006e-02	-0.351	0.725461	
## P_105	-2.783e-02	8.008e-02	-0.348	0.728233	
## P_106	-9.595e-04	8.144e-04	-1.178	0.238976	
## P_107	-4.662e-04	9.753e-04	-0.478	0.632724	
## P_108	6.670e-04	8.583e-04	0.777	0.437262	
## P_109	6.431e-05	2.303e-04	0.279	0.780146	
## P_110	3.128e-03	2.280e-03	1.372	0.170389	

```
## P_111
                              1.351e-04 3.240e-04 0.417 0.676742
## P_112
                              5.077e-04 3.210e-04 1.582 0.113956
## P_113
                             -2.719e-04 7.704e-05 -3.529 0.000434 ***
## P_114
                              3.860e-05 4.113e-04 0.094 0.925239
## P_115
                             -1.279e-04 2.321e-04 -0.551 0.581705
                             -1.893e-03 8.726e-04 -2.170 0.030248 *
## P_116
## P_117
                             -3.348e-04 1.170e-04 -2.862 0.004283 **
## P_118
                             1.024e-03 3.403e-04 3.007 0.002694 **
## P_119
                             -2.152e-04 2.094e-04 -1.028 0.304405
## P_120
                              2.477e-04 2.495e-04 0.993 0.320850
## P_121
                             -5.776e-04 8.697e-04 -0.664 0.506730
## mtrain
                                     NA
                                               NA
                                                       NA
                                                                NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.01974 on 1124 degrees of freedom
## Multiple R-squared: 0.8698, Adjusted R-squared: 0.8543
## F-statistic: 56.06 on 134 and 1124 DF, p-value: < 2.2e-16
rank_hat_ols = predict(to_hat, newdata=proj4)
## Warning in predict.lm(to_hat, newdata = proj4): prediction from a rank-
## deficient fit may be misleading
summary(rank_hat_ols); hist(rank_hat_ols, xlab="Predicted Rates - OLS")
##
     Min. 1st Qu. Median
                             Mean 3rd Qu.
                                            Max.
```

## 0.2339 0.3826 0.4130 0.4134 0.4422 0.6050

#### mologram or rame\_mat\_ors

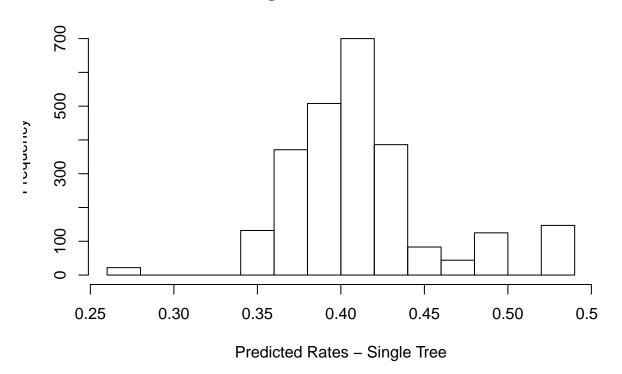


```
## n= 1259
##
## node), split, n, deviance, yval
## * denotes terminal node
##
## 1) root 1259 3.36343200 0.4132139
## 2) P_57>=14.75 805 1.35944700 0.3909387
## 4) P_37>=0.08508629 376 0.55982770 0.3676036
## 8) P_31>=0.260393 80 0.15583940 0.3312077
## 16) P_50>=0.9955153 12 0.02225725 0.2731211 *
```

```
17) P 50< 0.9955153 68 0.08594827 0.3414583 *
##
          9) P_31< 0.260393 296 0.26937450 0.3774403
##
##
           18) P_34>=0.1918283 183 0.10196010 0.3653738 *
           19) P 34< 0.1918283 113 0.09761885 0.3969817 *
##
##
        5) P_37< 0.08508629 429 0.41542780 0.4113910
         10) P_7>=38.91737 145 0.06953851 0.3934341 *
##
##
         11) P_7< 38.91737 284 0.27526220 0.4205592
           22) P_45>=0.2089347 261 0.20636630 0.4169893 *
##
           23) P 45< 0.2089347 23 0.02782549 0.4610691 *
##
##
      3) P_57< 14.75 454 0.89632250 0.4527106
##
        6) P_56>=3.05 323 0.32977590 0.4369945
         12) P_34>=0.005043071 289 0.22597590 0.4317835
##
##
           24) P_37>=0.03899463 92 0.08022539 0.4145569 *
           25) P_37< 0.03899463 197 0.10569890 0.4398285 *
##
         13) P_34< 0.005043071 34 0.02924967 0.4812873 *
##
##
        7) P_56< 3.05 131 0.29005690 0.4914611
         14) P_37>=0.006757407 70 0.09285862 0.4613694
##
           28) P 105>=4.94 42 0.03684633 0.4431274 *
##
           29) P_105< 4.94 28 0.02107163 0.4887323 *
##
##
         15) P_37< 0.006757407 61 0.06107474 0.5259926 *
rank_hat_tree <- predict(one_tree, newdata=proj4)</pre>
table(rank_hat_tree)
## rank_hat_tree
## 0.273121061424414 0.341458336833645 0.365373798406841 0.393434075445964
                                    132
                                                      371
                                                                         291
##
                  22
## 0.396981655496412 0.414556937049264 0.416989306152095 0.439828460591699
##
                 217
                                    167
                                                      533
                                                                         386
## 0.443127433458964 0.461069139449493 0.481287251500522 0.488732271960803
##
                  83
                                    44
                                                       65
                                                                          60
## 0.525992561070645
##
                 147
```

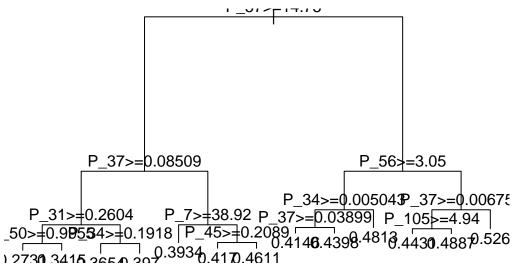
## hist(rank\_hat\_tree, xlab="Predicted Rates - Single Tree")

## mologram or rame\_mat\_tree



```
plot(one_tree) # plot tree

text(one_tree) # add labels to tree
```



# print complexity parameter table using cross validation
printcp(one\_tree)

##

```
## Regression tree:
## rpart(formula = reformulate(vars, "kfr_pooled_p25"), data = proj4,
##
       subset = training == 1, control = rpart.control(xval = 10))
##
## Variables actually used in tree construction:
## [1] P_105 P_31 P_34 P_37 P_45 P_50 P_56 P_57 P_7
##
## Root node error: 3.3634/1259 = 0.0026715
##
## n= 1259
##
##
            CP nsplit rel error xerror
                                            xstd
## 1 0.329325
                    0
                        1.00000 1.00182 0.049309
                        0.67068 0.67729 0.033871
## 2 0.114226
## 3 0.082205
                    2
                        0.55645 0.57380 0.029116
## 4 0.040472
                        0.47424 0.51889 0.026479
## 5 0.040023
                        0.43377 0.50260 0.024781
## 6 0.022165
                        0.39375 0.46136 0.022345
                    5
## 7 0.020999
                    6
                        0.37158 0.43949 0.021248
## 8 0.020751
                    7
                        0.35059 0.43623 0.021192
## 9 0.014162
                        0.32984 0.40788 0.020192
                    8
## 10 0.012211
                    9
                        0.31567 0.40360 0.019854
## 11 0.011908
                   10
                        0.30346 0.39973 0.019636
## 12 0.010388
                        0.29155 0.39204 0.019580
                   11
## 13 0.010000
                   12
                        0.28117 0.38971 0.019380
#Random Forest from 1000 Bootstrapped Samples
forest_hat <- randomForest(reformulate(vars, "kfr_pooled_p25"), ntree=1000, mtry=11, maxnodes=100</pre>
                           ,importance=TRUE, do.trace=25, data=proj4[proj4$training==1,], no.action = n
##
               Out-of-bag
## Tree |
               MSE %Var(y) |
     25 | 0.0006048
                       22.64 |
```

50 | 0.0005575

20.87

##

75 | 0.0005544 20.75 | ## 100 | 0.0005579 20.88 | ## ## 125 | 0.0005529 20.70 150 | 0.0005468 20.47 | ## ## 175 | 0.0005466 20.46 | 200 | 0.0005493 20.56 | ## ## 225 | 0.0005503 20.60 | ## 250 | 0.0005469 20.47 275 | 0.0005455 ## 20.42 | ## 300 | 0.0005459 20.43 | 325 | 0.0005456 20.42 | ## 350 | 0.0005441 ## 20.37 | ## 375 | 0.0005443 20.37 | 400 | 0.0005443 ## 20.38 | 425 | 0.0005451 20.40 | ## ## 450 | 0.0005433 20.34 | 475 | 0.0005435 20.35 | ## 500 | 0.000542 20.29 | ## 525 | 0.0005412 ## 20.26 ## 550 | 0.0005413 20.26 | 575 | 0.0005409 20.25 | ## 600 | 0.0005416 20.27 | ## ## 625 | 0.0005421 20.29 | 650 | 0.0005424 ## 20.30 | 675 | 0.0005422 20.29 | ## ## 700 | 0.0005426 20.31 | 725 | 0.0005419 20.28 | ## 750 | 0.000542 20.29 | ## ## 775 | 0.0005416 20.27 | 800 | 0.0005416 20.27 | ## 825 | 0.0005419 20.29 | ## 850 | 0.0005423 20.30 | 875 | 0.0005425 20.31 | ##

```
## 900 | 0.0005424 20.30 |

## 925 | 0.000543 20.33 |

## 950 | 0.0005425 20.31 |

## 975 | 0.0005425 20.31 |

## 1000 | 0.0005424 20.30 |
```

# getTree(forest\_hat, 250, labelVar = TRUE) #Text Representation of Tree

##		left daughter	right daughter	split var	split point
##	1	2	3	P_105	3.145000e+00
##	2	4	5	P_58	7.519521e+00
##	3	6	7	P_71	3.287916e+01
##	4	8	9	P_4	2.802782e+01
##	5	10	11	P_96	3.120000e+00
##	6	12	13	P_71	2.159720e+01
##	7	14	15	P_43	3.847482e-01
##	8	16	17	P_11	1.150000e+01
##	9	18	19	P_65	8.301119e+00
##	10	20	21	P_75	1.585000e+01
##	11	22	23	P_51	2.712052e+01
##	12	24	25	P_45	2.325134e-01
##	13	26	27	P_51	3.545973e+01
##	14	28	29	P_46	3.433064e-01
##	15	30	31	P_84	1.755000e+00
##	16	32	33	P_58	5.871207e+00
##	17	34	35	P_78	3.085000e+01
##	18	36	37	P_97	3.262000e+01
##	19	38	39	v14	-7.014916e-05
##	20	40	41	P_72	6.628010e+01
##	21	42	43	P_78	2.980000e+01
##	22	0	0	<na></na>	0.000000e+00
##	23	44	45	P_31	1.523578e-01
##	24	46	47	P_38	9.530012e-01
##	25	48	49	P_85	6.000000e+00

##	26	50	51	P_55	3.050000e+00
##	27	52	53	P_34	1.526280e-01
##	28	54	55	P_51	5.259372e+01
##	29	56	57	P_78	3.085000e+01
##	30	58	59	P_10	6.100000e+00
##	31	60	61	P_29	2.005410e+04
##	32	62	63	P_2	2.842626e+01
##	33	64	65	P_95	5.646000e+01
##	34	0	0	<na></na>	0.000000e+00
##	35	66	67	P_45	2.365560e-01
##	36	68	69	P_27	1.980439e-01
##	37	0	0	<na></na>	0.000000e+00
##	38	70	71	P_31	2.287695e-01
##	39	72	73	P_47	6.857073e-01
##	40	0	0	<na></na>	0.000000e+00
##	41	74	75	P_58	8.797444e+00
##	42	76	77	P_17	1.273779e+03
##	43	78	79	P_34	5.122802e-02
##	44	0	0	<na></na>	0.000000e+00
##	45	80	81	v14	-3.296223e-05
##	46	82	83	P_62	7.213700e+03
##	47	84	85	P_68	1.020000e+01
##	48	86	87	P_113	6.495000e+00
##	49	88	89 personageyears5c	nwardsnati	3.309562e-01
##	50	90	91	P_32	9.212385e-02
##	51	92	93	P_20	4.858952e+01
##	52	94	95	P_74	2.080000e+01
##	53	96	97	P_18	6.659283e+02
##	54	98	99	P_113	4.750000e+01
##	55 1	100	01 housingunitcashr	rentstatusw	7.398326e-02
##	56 1	102	03	P_117	1.068500e+01
##	57 1	104	05	P_57	1.365000e+01
##	58	0	0	<na></na>	0.000000e+00

##	59	106	107	P_56	4.550000e+00
##	60	108	109	P_54	2.040112e+02
##	61	110	111	P_45	3.659864e-01
##	62	112	113	P_55	2.400000e+00
##	63	114	115	P_81	8.400000e+00
##	64	116	117	P_104	3.200000e+00
##	65	118	119	P_48	8.637618e+01
##	66	0	0	<na></na>	0.000000e+00
##	67	0	0	<na></na>	0.000000e+00
##	68	0	0	<na></na>	0.000000e+00
##	69	120	121	personageyears18onwardsarm	7.284281e-02
##	70	122	123	v6	2.271584e-01
##	71	0	0	<na></na>	0.000000e+00
##	72	124	125	P_61	2.650089e+01
##	73	126	127	P_13	1.462570e-01
##	74	128	129	P_11	9.200000e+00
##	75	130	131	P_46	5.399068e-01
##	76	0	0	<na></na>	0.000000e+00
##	77	132	133	v14	-3.314143e-05
##	78	134	135	v13	4.494168e-03
##	79	136	137	P_55	4.450000e+00
##	80	138	139	personageyears25onwardsedu	2.815344e-03
##	81	0	0	<na></na>	0.000000e+00
##	82	0	0	<na></na>	0.000000e+00
##	83	140	141	P_103	1.220000e+00
##	84	142	143	v14	-2.965951e-05
##	85	0	0	<na></na>	0.000000e+00
##	86	0	0	<na></na>	0.000000e+00
##	87	144	145	P_78	2.810000e+01
##	88	146	147	P_18	1.878525e+02
##	89	0	0	<na></na>	0.000000e+00
##	90	148	149	P_43	3.201817e-01
##	91	0	0	<na></na>	0.000000e+00

##	92	0	0	<na></na>	0.000000e+00
##	93	150	151	personageyears25onwardsedu	2.539678e-03
##	94	152	153	P_91	6.100000e-01
##	95	154	155	P_56	4.200000e+00
##	96	156	157	P_84	8.345000e+00
##	97	158	159	P_17	2.957276e+03
##	98	160	161	P_94	4.150500e+01
##	99	162	163	P_41	2.238579e+00
##	100	0	0	<na></na>	0.000000e+00
##	101	164	165	P_96	4.215000e+00
##	102	166	167	v14	-6.321479e-05
##	103	168	169	v4	7.705185e-04
##	104	170	171	${\tt housing unit cash rent status w}$	2.412293e-01
##	105	172	173	P_15	6.314066e+01
##	106	174	175	P_4	1.093886e+01
##	107	176	177	P_40	2.086687e-03
##	108	0	0	<na></na>	0.000000e+00
##	109	178	179	P_50	9.809754e-01
##	110	180	181	P_4	1.862243e+01
##	111	182	183	P_29	2.980691e+04
##	112	0	0	<na></na>	0.000000e+00
##	113	0	0	<na></na>	0.000000e+00
##	114	184	185	P_60	3.725388e+01
##	115	0	0	<na></na>	0.000000e+00
##	116	186	187	P_41	4.137999e+00
##	117	188	189	P_97	9.905000e+00
##	118	0	0	<na></na>	0.000000e+00
##	119	0	0	<na></na>	0.000000e+00
##	120	0	0	<na></na>	0.000000e+00
##	121	0	0	<na></na>	0.000000e+00
##	122	0	0	<na></na>	0.000000e+00
##	123	190	191	P_6	2.632545e+01
##	124	0	0	<na></na>	0.000000e+00

##	125	192	193	P_41	2.897324e+00
##	126	0	0	<na></na>	0.000000e+00
##	127	0	0	<na></na>	0.000000e+00
##	128	194	195	P_39	2.404123e-02
##	129	196	197	P_33	1.801649e-01
##	130	198	199	P_79	3.110000e+01
##	131	0	0	<na></na>	0.000000e+00
##	132	0	0	<na></na>	0.000000e+00
##	133	0	0	<na></na>	0.000000e+00
##	134	0	0	<na></na>	0.000000e+00
##	135	0	0	<na></na>	0.000000e+00
##	136	0	0	<na></na>	0.000000e+00
##	137	0	0	<na></na>	0.000000e+00
##	138	0	0	<na></na>	0.000000e+00
##	139	0	0	<na></na>	0.000000e+00
##	140	0	0	<na></na>	0.000000e+00
##	141	0	0	<na></na>	0.000000e+00
##	142	0	0	<na></na>	0.000000e+00
##	143	0	0	<na></na>	0.000000e+00
##	144	0	0	<na></na>	0.000000e+00
##	145	0	0	<na></na>	0.000000e+00
##	146	0	0	<na></na>	0.000000e+00
##	147	0	0	<na></na>	0.000000e+00
##	148	0	0	<na></na>	0.000000e+00
##	149	0	0	<na></na>	0.000000e+00
##	150	0	0	<na></na>	0.000000e+00
##	151	0	0	<na></na>	0.000000e+00
##	152	0	0	<na></na>	0.000000e+00
##	153	0	0	<na></na>	0.000000e+00
##	154	0	0	<na></na>	0.000000e+00
##	155	0	0	<na></na>	0.000000e+00
##	156	0	0	<na></na>	0.000000e+00
##	157	0	0	<na></na>	0.000000e+00

##	158	0	0	<na></na>	0.000000e+00
##	159	0	0	<na></na>	0.000000e+00
##	160	0	0	<na></na>	0.000000e+00
##	161	0	0	<na></na>	0.000000e+00
##	162	0	0	<na></na>	0.000000e+00
##	163	0	0	<na></na>	0.000000e+00
##	164	0	0	<na></na>	0.000000e+00
##	165	0	0	<na></na>	0.000000e+00
##	166	0	0	<na></na>	0.000000e+00
##	167	0	0	<na></na>	0.000000e+00
##	168	0	0	<na></na>	0.000000e+00
##	169	0	0	<na></na>	0.000000e+00
##	170	0	0	<na></na>	0.000000e+00
##	171	0	0	<na></na>	0.000000e+00
##	172	0	0	<na></na>	0.000000e+00
##	173	0	0	<na></na>	0.000000e+00
##	174	0	0	<na></na>	0.000000e+00
##	175	0	0	<na></na>	0.000000e+00
##	176	0	0	<na></na>	0.000000e+00
##	177	0	0	<na></na>	0.000000e+00
##	178	0	0	<na></na>	0.000000e+00
##	179	0	0	<na></na>	0.000000e+00
##	180	0	0	<na></na>	0.000000e+00
##	181	0	0	<na></na>	0.000000e+00
##	182	0	0	<na></na>	0.000000e+00
##	183	0	0	<na></na>	0.000000e+00
##	184	0	0	<na></na>	0.000000e+00
##	185	0	0	<na></na>	0.000000e+00
##	186	0	0	<na></na>	0.000000e+00
##	187	0	0	<na></na>	0.000000e+00
##	188	0	0	<na></na>	0.000000e+00
##	189	0	0	<na></na>	0.000000e+00
##	190	0	0	<na></na>	0.000000e+00

191		0	0	<na></na>	0.000000e+00
192		0	0	<na></na>	0.000000e+00
193		0	0	<na></na>	0.000000e+00
194		0	0	<na></na>	0.000000e+00
195		0	0	<na></na>	0.000000e+00
196		0	0	<na></na>	0.000000e+00
197		0	0	<na></na>	0.000000e+00
198		0	0	<na></na>	0.000000e+00
199		0	0	<na></na>	0.000000e+00
	status	prediction			
1	-3	0.4150675			
2	-3	0.4596172			
3	-3	0.4050568			
4	-3	0.4888528			
5	-3	0.4162352			
6	-3	0.4328117			
7	-3	0.3917584			
8	-3	0.5223360			
9	-3	0.4686072			
10	-3	0.4289750			
11	-3	0.3592809			
12	-3	0.4500452			
13	-3	0.4196814			
14	-3	0.4035246			
15	-3	0.3597945			
16	-3	0.5382537			
17	-3	0.4462848			
18	-3	0.5060343			
19	-3	0.4631179			
20	-3	0.4034092			
21	-3	0.4545408			
22	-1	0.3897759			
	192 193 194 195 196 197 198 199 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21	192 193 194 195 196 197 198 199 status 1	192 0 193 0 194 0 195 0 196 0 197 0 198 0 199 0 199 0 status prediction 1 -3 0.4150675 2 -3 0.4596172 3 -3 0.4050568 4 -3 0.4888528 5 -3 0.4162352 6 -3 0.4328117 7 -3 0.3917584 8 -3 0.5223360 9 -3 0.4686072 10 -3 0.4289750 11 -3 0.3592809 12 -3 0.4500452 13 -3 0.4196814 14 -3 0.4035246 15 -3 0.3597945 16 -3 0.3597945 16 -3 0.5382537 17 -3 0.4462848 18 -3 0.5060343 19 -3 0.4631179 20 -3 0.4034092 21 -3 0.4545408	192 0 0 0 193 0 0 194 0 0 195 0 0 196 0 0 197 0 0 198 0 0 199 0 0 0 199 0 0 199 0 0 199 0 0 0 199 0 0 199 0 0 0 199 0 0 0 199 0 0 0 199 0 0 0 199 0 0 0 199 0 0 0 199 0 0 0 199 0 0 0	192 0 0 0 NA> 193 0 0 0 NA> 194 0 0 0 NA> 195 0 0 0 NA> 196 0 0 0 NA> 197 0 0 0 NA> 198 0 0 0 NA> 199 0 0 NA> 199 0 0 NA> 199 0 0 NA> 199 0 0 NA> 190 0 NA> 190 0 NA> 190 0 0 NA> 190

- ## 28 -3 0.3872964
- ## 29 -3 0.4172743
- ## 30 -3 0.3981197
- ## 31 -3 0.3509697
- ## 33 -3 0.5150889
- ## 35 -3 0.4519516
- **##** 36 -3 0.5242328
- ## 37 -1 0.4741869
- ## 38 -3 0.4366765
- ## 39 -3 0.4686620
- ## 40 -1 0.4907961
- ## 41 -3 0.3985544
- ## 42 -3 0.5012060
- ## 43 -3 0.4378747
- ## 44 -1 0.3845832
- ## 45 -3 0.3474402
- ## 46 -3 0.4570083
- ## 47 -3 0.5050129
- ## 48 -3 0.4087551
- **##** 49 -3 0.4432413
- ## 50 -3 0.4999307
- ## 51 -3 0.4373468
- ## 53 -3 0.3755459
- **##** 54 -3 0.4087126
- ## 55 -3 0.3797011
- ## 56 -3 0.4425626

## 57 -3 0.4061528 ## 58 -1 0.4605493 -3 0.3943361 ## 59 ## 60 -3 0.2840200 ## 61 -3 0.3572312 -3 0.5981257 ## 62 ## 63 -3 0.5427286 -3 0.5090275 ## 64 ## 65 -3 0.5279693 ## 66 -1 0.4579172 ## 67 -1 0.4420089 -1 0.5102413 ## 68 ## 69 -3 0.5265647 -3 0.4436183 ## 70 ## 71 -1 0.3984962 ## 72 -3 0.4643022 -3 0.4943365 ## 73 ## 74 -3 0.4110343 -3 0.3810824 ## 75 ## 76 -1 0.5575326 ## 77 -3 0.4871243 ## 78 -3 0.4655834 ## 79 -3 0.4247495 ## 80 -3 0.3428436 ## 81 -1 0.3704231 -1 0.5068583 ## 82 ## 83 -3 0.4526735 ## 84 -3 0.5010481 ## 85 -1 0.5684507 ## 86 -1 0.4419180 -3 0.3990013 ## 87 ## 88 -3 0.4402975

## 89

-1 0.4942672

## 90 -3 0.4770957 ## 91 -1 0.5318998 0.4718888 ## 92 ## 93 -3 0.4250104 ## 94 -3 0.4213671 -3 0.3803915 ## 95 ## 96 -3 0.3883214 -3 0.3499950 ## 97 ## 98 -3 0.4160090 ## 99 -3 0.3603739 ## 100 -1 0.2536505 ## 101 -3 0.3811841 ## 102 -3 0.4331327 -3 0.5051426 ## 103 ## 104 -3 0.4607956 ## 105 -3 0.4031339 -3 0.4027250 ## 106 ## 107 -3 0.3750415 -1 0.3540276 ## 108 ## 109 -3 0.2402652 ## 110 -3 0.3698732 ## 111 -3 0.3405858 ## 112 -1 0.6140298 -1 0.5854024 ## 113 ## 114 -3 0.5293418 -1 0.5895824 ## 115 ## 116 -3 0.5250566 ## 117 -3 0.4947794 ## 118 -1 0.5209529 ## 119 -1 0.5321792 ## 120 -1 0.5214394 ## 121 -1 0.5316899

## 122

-1 0.4610475

- ## 124 -1 0.5043333
- ## 125 -3 0.4619004
- ## 126 -1 0.5156768
- ## 127 -1 0.4676611
- ## 129 -3 0.4425528
- ## 130 -3 0.3717482
- ## 131 -1 0.4184196
- ## 132 -1 0.5005980
- ## 134 -1 0.4783471
- ## 135 -1 0.4400559
- ## 136 -1 0.4185162
- ## 137 -1 0.4481243
- ## 138 -1 0.3466686
- ## 139 -1 0.3412043
- ## 141 -1 0.4782005
- ## 142 -1 0.5125461
- ## 143 -1 0.4895501
- ## 144 -1 0.4067453
- ## 145 -1 0.3804155
- ## 146 -1 0.4753168
- ## 147 -1 0.4381535
- ## 149 -1 0.4302574
- ## 150 -1 0.4606226
- ## 151 -1 0.4190750
- ## 152 -1 0.4148664
- ## 153 -1 0.4334615
- ## 154 -1 0.3883879
- ## 155 -1 0.3564025

## 156 -1 0.3933553 ## 157 -1 0.3707028 ## 158 -1 0.3779216 ## 159 -1 0.3465041 ## 160 -1 0.4358945 -1 0.4095462 ## 161 ## 162 -1 0.3801639 -1 0.3484998 ## 163 -1 0.3863915 ## 164 ## 165 -1 0.3595654 ## 166 -1 0.4891859 -1 0.4242354 ## 167 ## 168 -1 0.4640645 -1 0.5205469 ## 169 ## 170 -1 0.4768224 ## 171 -1 0.4233997 -1 0.4078270 ## 172 ## 173 -1 0.3774895 -1 0.4624898 ## 174 ## 175 -1 0.4000084 ## 176 -1 0.4121799 ## 177 -1 0.3709150 ## 178 -1 0.2279315 -1 0.2420271 ## 179

## 180

## 181 ## 182

## 183 ## 184

## 185

## 186

## 187 ## 188 -1 0.4322645 -1 0.3656576

-1 0.3645755 -1 0.3363523

-1 0.4859396

-1 0.5365755

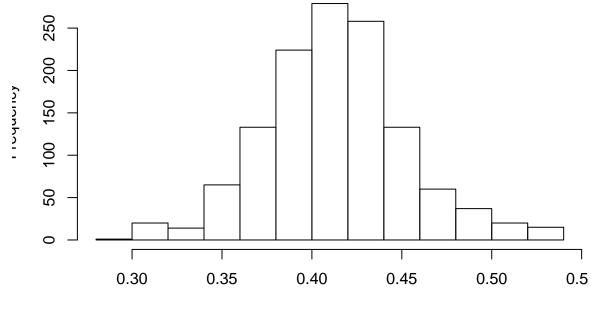
-1 0.5306628 -1 0.5157129

-1 0.4872797

24

```
## 189
           -1 0.4969222
## 190
               0.4359158
               0.4256831
## 191
## 192
           -1
               0.4403594
## 193
               0.4660034
               0.4033011
## 194
           -1
## 195
           -1
               0.3667450
## 196
               0.4650491
## 197
              0.4313046
           -1
           -1 0.3798913
## 198
## 199
           -1 0.3659316
```

```
rank_hat_forest <- predict(forest_hat, data=proj4,type="response")
summary(rank_hat_forest); hist(rank_hat_forest, xlab="Predicted Rates - Random Forest")</pre>
```

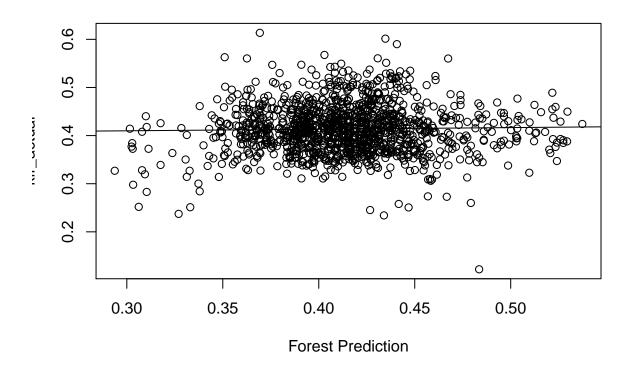


Predicted Rates - Random Forest

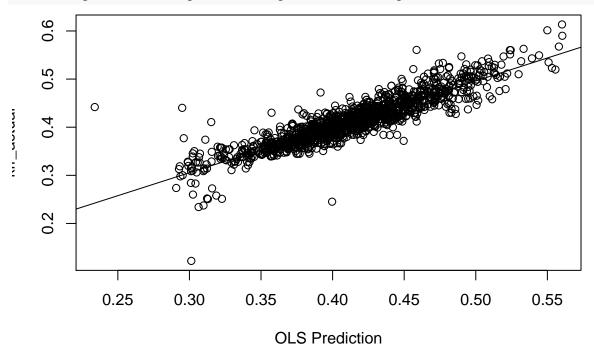
```
proj4$rank_hat_forest = rank_hat_forest
proj4$rank_hat_ols = rank_hat_ols
```

```
proj4$rank_hat_tree = rank_hat_tree
forest_pred_error <- proj4$kfr_pooled_p25[proj4$training == 1] - proj4$rank_hat_forest[proj4$training =
ols_pred_error <- proj4\$kfr_pooled_p25[proj4\$training == 1] - proj4\$rank_hat_ols[proj4\$training == 1]
tree_pred_error <- proj4$kfr_pooled_p25[proj4$training == 1] - proj4$rank_hat_tree[proj4$training == 1]
proj4$mse_forest = forest_pred_error^2
proj4$mse_ols = ols_pred_error^2
proj4$mse_tree = tree_pred_error^2
#mse for forest in sample
mean(proj4$mse_forest)
## [1] 0.004271803
mean(proj4$mse_ols)
## [1] 0.0003477129
mean(proj4$mse_tree)
## [1] 0.0007511369
#Export to stata
proj4$predictions_ols <- rank_hat_ols #Add OLS predictions to data set</pre>
proj4$predictions_tree <- rank_hat_tree #Add regression tree predictions to data set
proj4$predictions_forest <- rank_hat_forest #Add random forest predictions to data set
write.dta(proj4, "proj4_results.dta") #Save data as a stata .dta file
#project4$pred_error = project4$kfr_actual -project4$predictions
#project4$mse_forest = project4$pred_error^2
\#mse\_test \leftarrow subset(proj4, test==0, select = c(mse\_forest, mse\_trees, mse\_ols))
#summary(mse_test)
library("haven")
```

```
gdc<-read_dta("atlas_test.dta")</pre>
gdc<-merge(gdc,proj4,by = "geoid")</pre>
gdc$rank_hat_forest = rank_hat_forest
gdc$rank_hat_ols = rank_hat_ols
gdc$rank_hat_tree = rank_hat_tree
forest_pred_error <- gdc$kfr_actual[gdc$training == 0] - gdc$rank_hat_forest[gdc$training == 0]
ols_pred_error <- gdc$kfr_actual[gdc$training == 0] - gdc$rank_hat_ols[gdc$training == 0]
tree_pred_error <- gdc$kfr_actual[gdc$training == 0] - gdc$rank_hat_tree[gdc$training == 0]</pre>
gdc$mse_forest = forest_pred_error^2
gdc$mse_ols = ols_pred_error^2
gdc$mse_tree = tree_pred_error^2
#mse for forest in sample
mean(gdc$mse_forest)
## [1] 0.003974762
mean(gdc$mse_ols)
## [1] 0.0005073468
mean(gdc$mse_tree)
## [1] 0.001005525
gdc$predictions_ols <- rank_hat_ols #Add OLS predictions to data set</pre>
gdc$predictions_tree <- rank_hat_tree #Add regression tree predictions to data set
gdc$predictions_forest <- rank_hat_forest #Add random forest predictions to data set
write.dta(gdc, "gdc_results.dta") #Save data as a stata .dta file
plot(gdc$rank_hat_forest[gdc$test==1],gdc$kfr_actual[gdc$test==1], xlab = "Forest Prediction", ylab = "
abline(lm(gdc$kfr_actual[gdc$test==1]~gdc$rank_hat_forest[gdc$test==1]))
```



plot(gdc\$rank\_hat\_ols[gdc\$test==1],gdc\$kfr\_actual[gdc\$test==1], xlab = "OLS Prediction", ylab = "kfr\_ac
abline(lm(gdc\$kfr\_actual[gdc\$test==1]~gdc\$rank\_hat\_ols[gdc\$test==1]))



```
plot(gdc$rank_hat_tree[gdc$test==1],gdc$kfr_actual[gdc$test==1], xlab = "Tree Prediction", ylab = "kfr_
abline(lm(gdc$kfr_actual[gdc$test==1]~gdc$rank_hat_tree[gdc$test==1]))
```

