

ARE213 Problem Set #3

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1 Problem 1: Linear models

1.1 Part A: LM results comparison

Using a series of linear models (with heteroskedasticity consistent “robust” standard errors), we find that for a range of model formulations there is a significant effect on housing price from the presence of hazardous waste cleanup sites, but not in the direction one would expect. Instead of a reduction in housing value we estimate an increase in the value, which does not seem likely to be true. The coefficient for the hazardous waste indicator variable (npl2000) takes a wide range of values depending on which additional explanatory variables are included in the model, from 0.04 (i.e., approximately a 4% increase) for the simple model only including 1980 housing values and npl2000 to estimate 2000 housing values, to 0.09 for a model including both housing and demographic characteristics.

Requirements for Unbiased Estimates [add to this]: For our estimates to be unbiased we would need to include all of the potential sources of variation in housing price in a linear model. A particular challenge is that there are very few sites with NPL2000 status (only 2% of sites), so while the overall sample size is large there is very little support for estimates related to NPL2000 status compared to other covariates.

Table 1: Linear models for effect of NPL(2000) on housing value (with many additional state fixed effects omitted)

	simple model	+housing char.	+demographics	+state fixed effects
	(1)	(2)	(3)	(4)
npl2000	0.040*** (0.012)	0.055*** (0.012)	0.090*** (0.010)	0.068*** (0.009)
lnmeanhs8	0.856*** (0.011)	0.866*** (0.018)	0.619*** (0.022)	0.514*** (0.022)
firestoveheat80		0.074*** (0.020)	0.182*** (0.023)	0.230*** (0.033)
nofullkitchen80		-1.776*** (0.176)	-0.751*** (0.164)	-0.559*** (0.152)
zerofullbath80		1.243*** (0.139)	1.044*** (0.124)	0.863*** (0.116)
bedrms1_80occ		0.421* (0.249)	0.404* (0.237)	0.240 (0.234)
bedrms2_80occ		-0.436* (0.229)	0.156 (0.216)	-0.004 (0.214)
bedrms3_80occ		-0.524** (0.230)	-0.147 (0.217)	-0.153 (0.214)
bedrms4_80occ		-0.111 (0.226)	0.004 (0.217)	-0.213 (0.214)
bedrms5_80occ		0.721*** (0.231)	0.732*** (0.222)	0.430* (0.220)
blt0_1yrs80occ		-0.216*** (0.045)	-0.010 (0.044)	0.109** (0.045)
blt2_5yrs80occ		-0.295*** (0.029)	0.011 (0.028)	0.039 (0.026)
blt6_10yrs80occ		-0.271*** (0.021)	-0.048** (0.021)	0.002 (0.021)
blt10_20yrs80occ		-0.242*** (0.017)	-0.136*** (0.015)	-0.123*** (0.014)
blt20_30yrs80occ		-0.191*** (0.017)	-0.181*** (0.014)	-0.156*** (0.013)
blt30_40yrs80occ		-0.190*** (0.026)	-0.121*** (0.025)	-0.104*** (0.023)
occupied80		0.730*** (0.050)	0.242*** (0.046)	-0.093** (0.044)
pop_den8			0.00001*** (0.00000)	0.00001*** (0.00000)
shrbld8			-0.161*** (0.014)	-0.058*** (0.013)
shrhsp8			-0.329*** (0.021)	-0.100*** (0.022)
child8			-0.630*** (0.058)	-0.431*** (0.052)
old8			-0.737*** (0.047)	-0.447*** (0.044)
shrfor8			1.377*** (0.048)	0.567*** (0.041)
fth8			-0.006 (0.034)	-0.084*** (0.032)
smhse8			0.407*** (0.022)	0.323*** (0.022)
hsdrop8			0.010 (0.025)	0.042* (0.024)
no_hs_diploma8			-0.537*** (0.039)	-0.262*** (0.034)
ba_or_better8			0.112*** (0.034)	0.450*** (0.035)
unemp8			-0.654*** (0.071)	-1.420*** (0.076)
povrat8			-0.275*** (0.051)	0.118** (0.048)
welfare8			1.271*** (0.070)	0.284*** (0.067)
avhhin8			0.00001*** (0.00000)	0.00001*** (0.00000)
as.factor(statefips)2				-0.129*** (0.027)
as.factor(statefips)4				0.011 (0.015)
as.factor(statefips)5				-0.150*** (0.025)
as.factor(statefips)6				0.340*** (0.017)
as.factor(statefips)8				0.207*** (0.015)
as.factor(statefips)9				0.157*** (0.015)
as.factor(statefips)10				0.230*** (0.018)
as.factor(statefips)11				0.102*** (0.024)
as.factor(statefips)12				-0.005 (0.013)
as.factor(statefips)13				0.182*** (0.015)
as.factor(statefips)15				0.081** (0.038)
as.factor(statefips)16				0.039* (0.020)

Notes:

***Significant at the 1 percent level.

**Significant at the 5 percent level.

*Significant at the 10 percent level.

Table 2: Contingency table for a range of factors by nbr

	N	0			1		
		<i>N</i> = 42540			<i>N</i> = 5705		
npl1990 : 0	48245	99%	(42137)		94%	(5362)	
1		1%	(403)		6%	(343)	
npl2000 : 0	48245	99%	(41989)		92%	(5271)	
1		1%	(551)		8%	(434)	
pop_den8	48245	674.4158	2887.1943	6473.8845	194.1788	961.2832	3344.8530
shrbk8	48245	0.002638	0.015964	0.082081	0.001940	0.011711	0.058053
shrhsp8	48245	0.0064900	0.0185855	0.0634445	0.0049900	0.0131040	0.0369590
child8	48245	0.2328310	0.2846455	0.3289673	0.2587950	0.2974520	0.3323570
shrfor8	48245	0.01912050	0.04256600	0.08668875	0.01689000	0.03570600	0.06528300
ffh8	48245	0.09759375	0.15094300	0.24082675	0.08582000	0.12909600	0.19591101
smhse8	48245	0.4122337	0.5306680	0.6283828	0.4513610	0.5558160	0.6362780
hsdrop8	48245	0.054545	0.111739	0.192307	0.056338	0.107438	0.182572
no_hs_diploma8	48245	0.1816847	0.2916233	0.4290175	0.2093909	0.3010590	0.4139572
ba_or_better8	48245	0.07740766	0.13831653	0.24082869	0.07717157	0.12516019	0.20234250
unemp8	48245	0.03652375	0.05492750	0.08234525	0.04026800	0.05844100	0.08421600
povrat8	48245	0.0452915	0.0800220	0.1442960	0.0455540	0.0744300	0.1235710
welfare8	48245	0.0286895	0.0510035	0.0952380	0.0317910	0.0513860	0.0844040
favinc8	48245	18678.15	22858.37	27771.77	19537.96	22877.64	26884.79
avhhin8	48245	16223.85	20316.22	25271.48	17278.15	20688.77	24637.19
meanrnt80	48115	222.7207	268.2003	324.8214	224.3905	264.1910	313.5771
mdvalhs9	48245	43600	69000	125600	47300	72200	126112
meanrnt9	48190	389.6826	491.0426	635.9742	387.5005	487.7268	628.6454
mdvalhs0	48245	81900	120300	179700	87000	123100	167700
meanrnt0	48127	520.2832	645.7830	823.5361	520.2703	638.1877	800.3922
tothsun8	48245	873	1280	1739	891	1278	1718
ownocc8	48245	438	740	1080	544	831	1156
owner_occupied80	48245	0.4729324	0.6618154	0.7839525	0.5725154	0.7147436	0.8039216
bltlast5yrs80	48245	0.01945995	0.08701160	0.22249195	0.04517272	0.11816839	0.22092116
bltlast10yrs80	48245	0.06065490	0.21716790	0.45006267	0.12627292	0.26502535	0.42857143
firestoveheat80	48245	0.002498829	0.011919458	0.042651323	0.005777961	0.021126760	0.060995184
noaircond80	48245	0.1612903	0.3921197	0.6600715	0.2464638	0.4505289	0.6935139
nofullkitchen80	48245	0.004088785	0.009900990	0.020574207	0.004839685	0.010771993	0.020997375
zerofullbath80	48245	0.004034873	0.011940299	0.028044170	0.005586592	0.013818182	0.029411765
northeast : 0	48245	79%	(33483)		66%	(3779)	
1		3	21%	(9057)	34%	(1926)	
midwest : 0	48245	77%	(32659)		78%	(4449)	
1		23%	(9881)		22%	(1256)	
south : 0	48245	68%	(28851)		76%	(4327)	
1		32%	(13689)		24%	(1378)	
west : 0	48245	77%	(32627)		80%	(4560)	
1		23%	(9913)		20%	(1145)	

1.2 Part B: Comparing covariates

2 Problem 2: RDD setup

3 Problem 3: RDD First Stage

4 Problem 4: RDD Second Stage

5 Problem 5: Synthesis

6 Appendix: Code Listings

```
1 # Econometrics helper functions for [R]
2 #
3 # Peter Alstone and Frank Proulx
4 # 2013
5 # version 1
6 # contact: peter.alstone AT gmail.com
7
8 # Category: Data Management -----
9
10
11 # Category: Data Analysis -----
12
13 # Function: Find adjusted R^2 for subset of data
14 # This requires a completed linear model...pull out the relevant y-values
15 # and residuals and feed them to function
16 # [TODO @Peter] Improve function so it can simply evaluate lm or glm object,
17 # add error handling, general clean up.
18 adjr2 <- function(y,resid){
19   r2 <- 1-sum(resid^2) / sum((y-mean(y))^2)
20   return(r2)
21 } #end adjr2
22
23
24 # Category: Plots and Graphics -----
25
26 ## Function for arranging ggplots. use png(); arrange(p1, p2, ncol=1); dev.
27 off() to save.
28 require(grid)
29 vp.layout <- function(x, y) viewport(layout.pos.row=x, layout.pos.col=y)
30 arrange_ggplot2 <- function(..., nrow=NULL, ncol=NULL, as.table=FALSE) {
31   dots <- list(...)
32   n <- length(dots)
33   if(is.null(nrow) & is.null(ncol)) { nrow = floor(n/2) ; ncol = ceiling(n/
34     nrow)}
35   if(is.null(nrow)) { nrow = ceiling(n/ncol)}
36   if(is.null(ncol)) { ncol = ceiling(n/nrow)}
37   ## NOTE see n2mfrow in grDevices for possible alternative
```

```

34 grid.newpage()
35 pushViewport(viewport(layout=grid.layout(nrow,ncol) ))
36 ii.p <- 1
37 for(ii.row in seq(1, nrow)){
38   ii.table.row <- ii.row
39   if(as.table) {ii.table.row <- nrow - ii.table.row + 1}
40   for(ii.col in seq(1, ncol)){
41     ii.table <- ii.p
42     if(ii.p > n) break
43     print(dots[[ii.table]], vp=vp.layout(ii.table.row, ii.col))
44     ii.p <- ii.p + 1
45   }
46 }
47 }
48
49 robust <- function(model){ #This calculates the Huber-White Robust standard
  errors -- code from http://thetarzan.wordpress.com/2011/05/28/heteroskedasticity-robust-and-clustered-standard-errors-in-r/
50   s <- summary(model)
51   X <- model.matrix(model)
52   u2 <- residuals(model)^2
53   XDX <- 0
54
55   for(i in 1:nrow(X)) {
56     XDX <- XDX +u2[i]*X[i,]%*%t(X[i,])
57   }
58
59   # inverse(X'X)
60   XX1 <- solve(t(X)%*%X)
61
62   #Compute variance/covariance matrix
63   varcovar <- XX1 %*% XDX %*% XX1
64
65   # Degrees of freedom adjustment
66   dfc <- sqrt(nrow(X))/sqrt(nrow(X)-ncol(X))
67
68   stdh <- dfc*sqrt(diag(varcovar))
69
70   t <- model$coefficients/stdh
71   p <- 2*pnorm(-abs(t))
72   results <- cbind(model$coefficients, stdh, t, p)
73   dimnames(results) <- dimnames(s$coefficients)
74   results
75 }
76
77 ## Two functions for clustered standard errors below from: http://people.su.se/~ma/clustering.pdf -----
78
79 clx <-
80 function(fm, dfcw, cluster){
81   # R-codes (www.r-project.org) for computing
82   # clustered-standard errors. Mahmood Arai, Jan 26, 2008.
83
84   # The arguments of the function are:
85   # fitted model, cluster1 and cluster2
86   # You need to install libraries 'sandwich' and 'lmtest'
87

```

```

88 # reweighting the var-cov matrix for the within model
89 library(sandwich);library(lmtest)
90 M <- length(unique(cluster))
91 N <- length(cluster)
92 K <- fm$rank
93 dfc <- (M/(M-1))*((N-1)/(N-K))
94 uj <- apply(estfun(fm),2, function(x) tapply(x, cluster, sum));
95 vcovCL <- dfc*sandwich(fm, meat=crossprod(uj)/N)*dfcw
96 coeftest(fm, vcovCL) }
97
98 mclx <-
99 function(fm, dfcw, cluster1, cluster2){
100 # R-codes (www.r-project.org) for computing multi-way
101 # clustered-standard errors. Mahmood Arai, Jan 26, 2008.
102 # See: Thompson (2006), Cameron, Gelbach and Miller (2006)
103 # and Petersen (2006).
104 # reweighting the var-cov matrix for the within model
105
106 # The arguments of the function are:
107 # fitted model, cluster1 and cluster2
108 # You need to install libraries 'sandwich' and 'lmtest'
109
110 library(sandwich);library(lmtest)
111 cluster12 = paste(cluster1,cluster2, sep=" ")
112 M1 <- length(unique(cluster1))
113 M2 <- length(unique(cluster2))
114 M12 <- length(unique(cluster12))
115 N <- length(cluster1)
116 K <- fm$rank
117 dfc1 <- (M1/(M1-1))*((N-1)/(N-K))
118 dfc2 <- (M2/(M2-1))*((N-1)/(N-K))
119 dfc12 <- (M12/(M12-1))*((N-1)/(N-K))
120 u1j <- apply(estfun(fm), 2, function(x) tapply(x, cluster1, sum))
121 u2j <- apply(estfun(fm), 2, function(x) tapply(x, cluster2, sum))
122 u12j <- apply(estfun(fm), 2, function(x) tapply(x, cluster12, sum))
123 vc1 <- dfc1*sandwich(fm, meat=crossprod(u1j)/N )
124 vc2 <- dfc2*sandwich(fm, meat=crossprod(u2j)/N )
125 vc12 <- dfc12*sandwich(fm, meat=crossprod(u12j)/N)
126 vcovMCL <- (vc1 + vc2 - vc12)*dfcw
127 coeftest(fm, vcovMCL)}
128
129 ## Function to compute ols standard errors , robust, clustered...
130 ## Based on http://diffuseprior.wordpress.com/2012/06/15/standard-robust-and-clustered-standard-errors-computed-in-r/
131 ols.hetero <- function(form, data, robust=FALSE, cluster=NULL,digits=3){
132 r1 <- lm(form, data)
133 if(length(cluster)!=0){
134 data <- na.omit(data[,c(colnames(r1$model),cluster)])
135 r1 <- lm(form, data)
136 }
137 X <- model.matrix(r1)
138 n <- dim(X)[1]
139 k <- dim(X)[2]
140 if(robust==FALSE & length(cluster)==0){
141 se <- sqrt(diag(solve(crossprod(X)) * as.numeric(crossprod(resid(r1))/(n-k))))
142 res <- cbind(coef(r1),se)

```

```

143 }
144 if(robust==TRUE){
145   u <- matrix(resid(r1))
146   meat1 <- t(X) %*% diag(diag(crossprod(t(u)))) %*% X
147   dfc <- n/(n-k)
148   se <- sqrt(dfc*diag(solve(crossprod(X)) %*% meat1 %*% solve(crossprod(X)
149   )))
149   res <- cbind(coef(r1),se)
150 }
151 if(length(cluster)!=0){
152   clus <- cbind(X,data[,cluster],resid(r1))
153   colnames(clus)[(dim(clus)[2]-1):dim(clus)[2]] <- c(cluster,"resid")
154   m <- dim(table(clus[,cluster]))
155   dfc <- (m/(m-1))*((n-1)/(n-k))
156   uclust <- apply(resid(r1)*X,2, function(x) tapply(x, clus[,cluster],
157   sum))
157   se <- sqrt(diag(solve(crossprod(X)) %*% (t(uclust) %*% uclust) %*% solve
158   (crossprod(X))*dfc)
158   res <- cbind(coef(r1),se)
159 }
160 res <- cbind(res,res[,1]/res[,2],(1-pnorm(abs(res[,1]/res[,2])))*2)
161 res1 <- matrix(as.numeric(sprintf(paste("%. ",paste(digits,"f",sep=""),sep=
162   " "),res)),nrow=dim(res)[1])
162 rownames(res1) <- rownames(res)
163 colnames(res1) <- c("Estimate","Std. Error","t value","Pr(>|t|)")
164 return(res1)
165 }

```

../util/are213-func.R