

Problem Set 1

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```
## Q1
```

```
library(foreign)
```

```
# Read in data
```

```
# taxi <- read.dta("./Spring 2017/Research Methods/02 taxi_phd.dta")
```

```
taxi <- read.dta("./02 taxi_phd.dta")
```

```
taxi_no_nas <- na.omit(taxi[,-2])
```

```
head(taxi)
```

```
##   fips tcrpid taxis   downrt  msa medianinc89 worktransit evercmptr
## 1 1097    NA    25 20.00000 5160      27601        1.2         0
## 2 1097    370    40 19.73684 5160      27601        1.2         0
## 3 2220    170     3  0.00000 9999      49327        1.6         0
## 4 2261    634     3  0.00000 9999      52929        0.7         0
## 5 4013    NA    550 0.00000 6200      36078        2.1         1
## 6 4013    408    15 50.00000 6200      36078        2.1         0
##   logtaxis logtaxisq logdensity northeast midwestdum southdum
## 1 3.218876 10.361162  7.430796          0           0         1
## 2 3.688879 13.607832  7.430796          0           0         1
## 3 1.098612  1.206949  8.000000          0           0         0
## 4 1.098612  1.206949  2.900467          0           0         0
## 5 6.309918 39.815071  7.930913          0           0         0
## 6 2.708050  7.333536  8.061882          0           0         0
```

```
summary(taxi)
```

```
##           fips           tcrpid           taxis           downrt
## Min.   : 1097   Min.   :  2.0   Min.   :  1.00   Min.   :  0.00
## 1st Qu.:15003   1st Qu.:198.0   1st Qu.:   7.00   1st Qu.:  0.00
## Median :29147   Median :358.5   Median :  19.00   Median :  0.00
## Mean   :28802   Mean   :342.5   Mean   :  63.98   Mean   : 20.34
## 3rd Qu.:41007   3rd Qu.:476.8   3rd Qu.:  50.00   3rd Qu.: 29.85
## Max.   :56021   Max.   :676.0   Max.   :1484.00   Max.   :100.00
## NA's   :1       NA's   :267
##           msa           medianinc89           worktransit           evercmptr
## Min.   :    0   Min.   :19527   Min.   : 0.000   Min.   :0.0000
## 1st Qu.:2160   1st Qu.:32214   1st Qu.: 0.800   1st Qu.:0.0000
## Median :5380   Median :36330   Median : 2.200   Median :0.0000
## Mean   :5027   Mean   :38252   Mean   : 5.244   Mean   :0.3064
## 3rd Qu.:7320   3rd Qu.:42634   3rd Qu.: 6.500   3rd Qu.:1.0000
## Max.   :9999   Max.   :62749   Max.   :58.400   Max.   :1.0000
## NA's   :1       NA's   :5
##           logtaxis           logtaxisq           logdensity           northeast
## Min.   :0.000   Min.   : 0.000   Min.   : 2.900   Min.   :0.0000
```

```
## 1st Qu.:1.946 1st Qu.: 3.787 1st Qu.: 7.591 1st Qu.:0.0000
## Median :2.944 Median : 8.670 Median : 8.000 Median :0.0000
## Mean :3.023 Mean :11.277 Mean : 8.013 Mean :0.2857
## 3rd Qu.:3.912 3rd Qu.:15.304 3rd Qu.: 8.381 3rd Qu.:1.0000
## Max. :7.302 Max. :53.326 Max. :10.712 Max. :1.0000
##
## midwestdum southdum
## Min. :0.0000 Min. :0.0000
## 1st Qu.:0.0000 1st Qu.:0.0000
## Median :0.0000 Median :0.0000
## Mean :0.2186 Mean :0.1824
## 3rd Qu.:0.0000 3rd Qu.:0.0000
## Max. :1.0000 Max. :1.0000
##
```

```
cor(taxi$evercmptr, taxi$logtaxis)
```

```
## [1] 0.5287976
```

```
cor(taxi$evercmptr, taxi$logtaxisq)
```

```
## [1] 0.5484535
```

```
cor(taxi_no_nas$evercmptr, taxi_no_nas$msa)
```

```
## [1] -0.07315375
```

```
cor(taxi_no_nas)
```

```
##          fips      taxis      downrt      msa medianinc89
## fips      1.00000000 -0.1231353607 -0.033433544 -0.01330391 -0.12142977
## taxis    -0.12313536  1.0000000000  0.144118182 -0.06460424  0.06195537
## downrt   -0.03343354  0.1441181821  1.000000000  0.03493522  0.05192196
## msa      -0.01330391 -0.0646042448  0.034935221  1.00000000 -0.05844503
## medianinc89 -0.12142977 0.0619553694 0.051921962 -0.05844503 1.00000000
## worktransit -0.10156608 0.2445882522 0.185140015 -0.04187248 0.26751690
## evercmptr  -0.07423793 0.4306553816 0.009279573 -0.07315375 0.06615210
## logtaxis   -0.06742299 0.7013789568 0.135538282 -0.12323772 0.16732118
## logtaxisq  -0.10754376 0.8353026288 0.173164138 -0.09765490 0.14065099
## logdensity -0.03556574 0.2599082773 0.185076927 -0.14703670 0.31980268
## northeast  0.11338316 -0.0890843309 -0.060683507 -0.10701046 0.42203041
## midwestdum 0.15059653 -0.0006121092 -0.062162470 -0.07448855 -0.19832548
## southdum   -0.03267988 0.0289440993 0.008585447  0.05130345 -0.21253221
##          worktransit evercmptr logtaxis logtaxisq logdensity
## fips      -0.10156608 -0.074237928 -0.06742299 -0.10754376 -0.03556574
## taxis      0.24458825 0.430655382  0.70137896  0.83530263  0.25990828
## downrt     0.18514001 0.009279573  0.13553828  0.17316414  0.18507693
## msa       -0.04187248 -0.073153753 -0.12323772 -0.09765490 -0.14703670
## medianinc89 0.26751690 0.066152099 0.16732118 0.14065099 0.31980268
## worktransit 1.00000000 0.181845756 0.30404935 0.31707935 0.51463528
## evercmptr   0.18184576 1.000000000 0.52915519 0.55050439 0.20726383
```

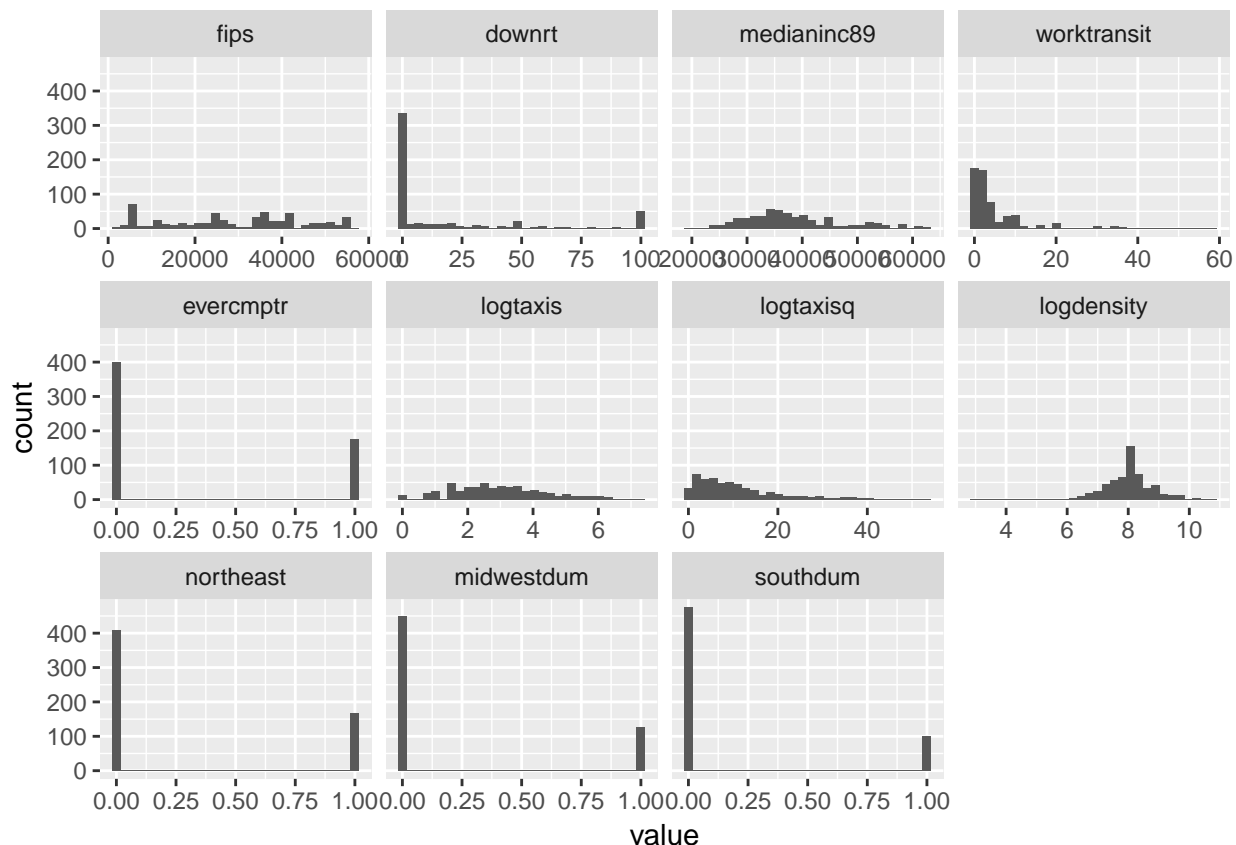
```
## logtaxi      0.30404935  0.529155186  1.00000000  0.96127516  0.37520013
## logtaxisq    0.31707935  0.550504388  0.96127516  1.00000000  0.36977535
## logdensity   0.51463528  0.207263832  0.37520013  0.36977535  1.00000000
## northeast    0.31596320 -0.136971218 -0.01567326 -0.05481631  0.25556704
## midwestdum  -0.12807643  0.012883723 -0.07286115 -0.06178894 -0.01930884
## southdum     -0.16655781  0.003119463  0.09425321  0.08121722 -0.23731592
##              northeast  midwestdum  southdum
## fips          0.11338316  0.1505965326 -0.032679884
## taxi         -0.08908433 -0.0006121092  0.028944099
## downrt       -0.06068351 -0.0621624700  0.008585447
## msa          -0.10701046 -0.0744885540  0.051303450
## medianinc89  0.42203041 -0.1983254788 -0.212532209
## worktransit  0.31596320 -0.1280764312 -0.166557808
## evercmptr    -0.13697122  0.0128837226  0.003119463
## logtaxi      -0.01567326 -0.0728611471  0.094253205
## logtaxisq    -0.05481631 -0.0617889429  0.081217223
## logdensity   0.25556704 -0.0193088375 -0.237315925
## northeast    1.00000000 -0.3384081437 -0.293410734
## midwestdum   -0.33840814  1.0000000000 -0.245240714
## southdum     -0.29341073 -0.2452407141  1.000000000
```

```
library(reshape2)
library(ggplot2)
d <- melt(taxi_no_nas[,c(1,3,5,6,7,8,9,10,11,12,13)])
```

```
## No id variables; using all as measure variables
```

```
ggplot(d,aes(x = value)) +
  facet_wrap(~variable,scales = "free_x") +
  geom_histogram()
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



Q1) The correlations between evercmptr and logtaxis and logtaxisq are 0.529 and 0.551 respectively, suggesting that computer adoption and firm size are positively correlated. The correlation between evercmptr logdensity is 0.207, suggesting that computer adoption and population density are positively correlated. The correlation between evercmptr and northeast, midwest, south dummy variables are -0.137, 0.013 and 0.003 respectively.

Q2

```
logit <- lm(evercmptr ~ .-evercmptr -taxis, data = taxi_no_nas, family = "binomial")
```

```
## Warning: In lm.fit(x, y, offset = offset, singular.ok = singular.ok, ...) :  
## extra argument 'family' will be disregarded
```

```
summary(logit)
```

```
##  
## Call:  
## lm(formula = evercmptr ~ . - evercmptr - taxis, data = taxi_no_nas,  
##     family = "binomial")  
##  
## Residuals:  
##      Min       1Q   Median       3Q      Max   
## -0.8553 -0.2395 -0.1020  0.1461  1.0671   
##  
## Coefficients:  
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) -9.271e-03  2.127e-01  -0.044  0.965245
## fips         3.875e-07  1.083e-06   0.358  0.720653
## downrt      -1.538e-03  5.068e-04  -3.034  0.002522 **
## msa         -4.306e-06  5.272e-06  -0.817  0.414352
## medianinc89  1.680e-06  2.124e-06   0.791  0.429242
## worktransit  3.404e-03  2.429e-03   1.401  0.161730
## logtaxis     8.305e-03  4.088e-02   0.203  0.839095
## logtaxisq    2.358e-02  6.043e-03   3.902  0.000107 ***
## logdensity   8.145e-03  2.617e-02   0.311  0.755720
## northeast   -1.952e-01  4.676e-02  -4.175  3.46e-05 ***
## midwestdum  -3.981e-02  4.566e-02  -0.872  0.383599
## southdum    -1.004e-01  4.876e-02  -2.059  0.039937 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3789 on 564 degrees of freedom
## Multiple R-squared:  0.3353, Adjusted R-squared:  0.3223
## F-statistic: 25.86 on 11 and 564 DF,  p-value: < 2.2e-16
```

Q2) From the regression model, we can see that downrt, northeast, and south have significant coefficients. The coefficients suggest that the log odds ratio of adopting computer technology associated with a 1% increase in driver ownership is about -1.54e-03, that the log odds ratio of adopting computer technology associated with being in northeast is about -1.95e-01, and that the log odds ratio of adopting computer technology associated with being in south is about -1.04e-01.

```
## Q3
```

```
# a)
# Add control
treatment <- lm(downrt ~ evercmptr + logtaxis + logtaxisq, data = taxi_no_nas)
summary(treatment)
```

```
##
## Call:
## lm(formula = downrt ~ evercmptr + logtaxis + logtaxisq, data = taxi_no_nas)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -44.902 -18.123 -16.787   5.753  89.595
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   26.7611     5.1325   5.214 2.59e-07 ***
## evercmptr     -8.7761     3.4636  -2.534  0.0115 *
## logtaxis      -9.1227     3.3077  -2.758  0.0060 **
## logtaxisq      2.0860     0.4948   4.216 2.90e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 31.92 on 572 degrees of freedom
## Multiple R-squared:  0.0532, Adjusted R-squared:  0.04824
## F-statistic: 10.71 on 3 and 572 DF,  p-value: 7.359e-07
```

```

# b)
# Robust SE
library(sandwich)
library(lmtest)

## Loading required package: zoo

##
## Attaching package: 'zoo'

## The following objects are masked from 'package:base':
##
##      as.Date, as.Date.numeric

coeftest(treatment, vcov = vcovHC(treatment, "HC1"))

##
## t test of coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 26.76106    6.82296  3.9222 9.839e-05 ***
## evercmptr   -8.77612    3.13380 -2.8005 0.0052752 **
## logtaxis    -9.12273    4.21489 -2.1644 0.0308464 *
## logtaxisq    2.08599    0.58096  3.5906 0.0003583 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# Clustered SE
# compute Stata like df-adjustment
G <- length(unique(taxi_no_nas$fips))
N <- length(taxi_no_nas$fips)
dfa <- (G/(G - 1)) * (N - 1)/treatment$df.residual
# display with cluster VCE and df-adjustment
country_vcov <- dfa * vcovHC(treatment, type = "HCO", cluster = "group", adjust = T)
coeftest(treatment, vcov = country_vcov)

##
## t test of coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 26.76106    6.82771  3.9195 9.949e-05 ***
## evercmptr   -8.77612    3.13598 -2.7985 0.0053067 **
## logtaxis    -9.12273    4.21783 -2.1629 0.0309627 *
## logtaxisq    2.08599    0.58136  3.5881 0.0003617 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

# c)
# Fixed Effects of MSA
treatment_msa <- lm(downrt ~ factor(msa) + evercmptr + logtaxis + logtaxisq, data = taxi_no_nas)
summary(treatment_msa)

```

```
##
## Call:
## lm(formula = downrt ~ factor(msa) + evercmptr + logtaxis + logtaxisq,
##     data = taxi_no_nas)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -57.774 -12.240  -0.969   3.138  97.626
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    16.284401     7.970053   2.043  0.041680 *
## factor(msa)80     0.424847    31.317905   0.014  0.989183
## factor(msa)160    19.344142    16.337206   1.184  0.237086
## factor(msa)200     0.126669    31.333625   0.004  0.996776
## factor(msa)220     0.465207    31.272697   0.015  0.988139
## factor(msa)240   -10.079845    31.108108  -0.324  0.746085
## factor(msa)280     2.911408    31.072297   0.094  0.925396
## factor(msa)460     0.464176    22.603687   0.021  0.983626
## factor(msa)520    19.846804    13.101528   1.515  0.130591
## factor(msa)560     4.218403    16.280834   0.259  0.795688
## factor(msa)600    24.096704    22.572924   1.068  0.286380
## factor(msa)640    48.179157    18.985266   2.538  0.011532 *
## factor(msa)720    12.998090    12.928461   1.005  0.315311
## factor(msa)730     0.600180    22.607511   0.027  0.978833
## factor(msa)760     3.196467    31.257115   0.102  0.918598
## factor(msa)870    -9.518157    31.082845  -0.306  0.759595
## factor(msa)875    -8.511729    16.296823  -0.522  0.601751
## factor(msa)1010  -11.126842    31.138026  -0.357  0.721024
## factor(msa)1120   -3.715275     8.901891  -0.417  0.676638
## factor(msa)1260   21.878263    31.144842   0.702  0.482791
## factor(msa)1280    9.283253    16.342610   0.568  0.570323
## factor(msa)1305  -10.236697    31.113179  -0.329  0.742315
## factor(msa)1400  -10.195163    31.085042  -0.328  0.743099
## factor(msa)1440   14.917001    31.256088   0.477  0.633441
## factor(msa)1480  -10.884517    31.131660  -0.350  0.726799
## factor(msa)1520    6.568331    18.565092   0.354  0.723673
## factor(msa)1540   31.133557    22.457098   1.386  0.166401
## factor(msa)1580  -10.195163    31.085042  -0.328  0.743099
## factor(msa)1600   26.294442    11.438957   2.299  0.022032 *
## factor(msa)1640   36.387408    11.709949   3.107  0.002020 **
## factor(msa)1760   15.083255    31.256948   0.483  0.629671
## factor(msa)1840   19.093520    16.350461   1.168  0.243588
## factor(msa)1920   66.763254    18.895713   3.533  0.000458 ***
## factor(msa)1960  -12.900505    31.245657  -0.413  0.679918
## factor(msa)2000   -0.866381    22.580151  -0.038  0.969412
## factor(msa)2080   55.007789    16.902772   3.254  0.001232 **
## factor(msa)2160   16.201016     9.963173   1.626  0.104708
## factor(msa)2240  -11.210043    22.377367  -0.501  0.616675
## factor(msa)2320   67.009437    31.185486   2.149  0.032247 *
## factor(msa)2400   -6.492459    18.550872  -0.350  0.726535
## factor(msa)2520   -5.155625    22.392018  -0.230  0.818019
## factor(msa)2560    0.476998    31.199730   0.015  0.987810
## factor(msa)2680   13.727898    23.287744   0.589  0.555861
```

```

## factor(msa)2720 37.717821 31.167269 1.210 0.226918
## factor(msa)2760 -9.707061 31.072977 -0.312 0.754901
## factor(msa)2840 -11.455069 31.144842 -0.368 0.713215
## factor(msa)2900 -0.075184 31.264916 -0.002 0.998082
## factor(msa)2920 -10.195163 31.085042 -0.328 0.743099
## factor(msa)2975 -9.605237 31.088671 -0.309 0.757509
## factor(msa)2995 -9.553375 22.329498 -0.428 0.668998
## factor(msa)3000 56.112706 31.181888 1.800 0.072679 .
## factor(msa)3060 0.241988 31.269056 0.008 0.993829
## factor(msa)3080 -10.195163 31.085042 -0.328 0.743099
## factor(msa)3120 31.981947 18.532781 1.726 0.085165 .
## factor(msa)3150 -1.255259 31.072297 -0.040 0.967796
## factor(msa)3160 22.367775 31.133809 0.718 0.472900
## factor(msa)3280 -5.340784 31.260781 -0.171 0.864430
## factor(msa)3320 12.696324 15.140994 0.839 0.402222
## factor(msa)3360 39.878405 22.462436 1.775 0.076593 .
## factor(msa)3400 -10.640533 22.398819 -0.475 0.635008
## factor(msa)3480 -8.929173 18.674315 -0.478 0.632800
## factor(msa)3520 -9.707061 31.072977 -0.312 0.754901
## factor(msa)3560 -9.490862 31.080022 -0.305 0.760242
## factor(msa)3600 -4.594232 18.574667 -0.247 0.804772
## factor(msa)3605 89.763303 31.113179 2.885 0.004123 **
## factor(msa)3610 -12.995731 22.516863 -0.577 0.564155
## factor(msa)3620 -10.195163 31.085042 -0.328 0.743099
## factor(msa)3640 -7.587954 14.772603 -0.514 0.607777
## factor(msa)3660 -10.195163 31.085042 -0.328 0.743099
## factor(msa)3760 55.386396 22.408590 2.472 0.013859 *
## factor(msa)3800 -9.896986 31.076444 -0.318 0.750291
## factor(msa)3810 -9.588592 31.072297 -0.309 0.757792
## factor(msa)3850 -9.605237 31.088671 -0.309 0.757509
## factor(msa)3980 -9.896986 31.076444 -0.318 0.750291
## factor(msa)4000 -14.027673 31.213816 -0.449 0.653379
## factor(msa)4040 0.005291 31.265887 0.000 0.999865
## factor(msa)4080 -9.605237 31.088671 -0.309 0.757509
## factor(msa)4100 -9.612609 22.327516 -0.431 0.667041
## factor(msa)4120 -13.670117 16.428778 -0.832 0.405852
## factor(msa)4200 23.854716 31.077365 0.768 0.443177
## factor(msa)4240 7.148509 31.082845 0.230 0.818222
## factor(msa)4280 -4.243591 31.252447 -0.136 0.892059
## factor(msa)4320 -10.236697 31.113179 -0.329 0.742315
## factor(msa)4400 -4.891515 31.165409 -0.157 0.875360
## factor(msa)4480 1.767368 11.327737 0.156 0.876094
## factor(msa)4520 -6.792289 22.735088 -0.299 0.765278
## factor(msa)4560 -9.490862 31.080022 -0.305 0.760242
## factor(msa)4600 -10.157708 31.110670 -0.327 0.744213
## factor(msa)4720 -6.967245 22.414783 -0.311 0.756088
## factor(msa)4900 -11.288861 22.338231 -0.505 0.613581
## factor(msa)4920 29.627492 31.250214 0.948 0.343658
## factor(msa)5015 33.530276 13.691905 2.449 0.014752 *
## factor(msa)5080 -0.486916 12.324185 -0.040 0.968504
## factor(msa)5120 -9.958299 22.345828 -0.446 0.656092
## factor(msa)5140 0.614772 31.307602 0.020 0.984343
## factor(msa)5160 8.948322 22.406125 0.399 0.689832
## factor(msa)5190 21.282168 18.538992 1.148 0.251659

```



```

## factor(msa)5200 0.465207 31.272697 0.015 0.988139
## factor(msa)5330 -10.668569 31.103768 -0.343 0.731777
## factor(msa)5345 -9.490862 31.080022 -0.305 0.760242
## factor(msa)5380 -3.191774 10.604272 -0.301 0.763577
## factor(msa)5400 -1.270511 22.339514 -0.057 0.954675
## factor(msa)5480 1.196967 22.369538 0.054 0.957353
## factor(msa)5560 31.073725 18.696843 1.662 0.097291 .
## factor(msa)5600 41.489552 10.876540 3.815 0.000158 ***
## factor(msa)5640 4.424069 13.813544 0.320 0.748929
## factor(msa)5660 -9.505222 22.330393 -0.426 0.670580
## factor(msa)5720 -10.218472 13.828429 -0.739 0.460367
## factor(msa)5775 23.789338 10.192060 2.334 0.020078 *
## factor(msa)5800 -9.112704 18.534493 -0.492 0.623225
## factor(msa)5920 -5.077944 22.583745 -0.225 0.822210
## factor(msa)5945 -6.089332 31.269477 -0.195 0.845696
## factor(msa)6080 -1.238671 22.344039 -0.055 0.955818
## factor(msa)6120 -5.841000 22.388877 -0.261 0.794311
## factor(msa)6160 4.781868 12.244719 0.391 0.696354
## factor(msa)6200 15.254947 12.336787 1.237 0.216974
## factor(msa)6280 -11.530757 18.563669 -0.621 0.534853
## factor(msa)6320 -9.557124 31.085746 -0.307 0.758664
## factor(msa)6340 0.424847 31.317905 0.014 0.989183
## factor(msa)6400 -9.525769 18.523380 -0.514 0.607353
## factor(msa)6440 31.198837 9.661068 3.229 0.001342 **
## factor(msa)6480 10.197977 22.328419 0.457 0.648113
## factor(msa)6640 44.260565 14.790047 2.993 0.002936 **
## factor(msa)6680 -10.984777 18.578709 -0.591 0.554679
## factor(msa)6760 -3.520278 22.383272 -0.157 0.875108
## factor(msa)6780 6.686464 14.838531 0.451 0.652508
## factor(msa)6800 -7.948482 22.394713 -0.355 0.722829
## factor(msa)6840 89.460909 22.379393 3.997 7.61e-05 ***
## factor(msa)6920 40.134531 12.931179 3.104 0.002045 **
## factor(msa)6960 -9.518157 22.340248 -0.426 0.670295
## factor(msa)7040 -3.903501 13.725501 -0.284 0.776251
## factor(msa)7080 -10.079845 31.108108 -0.324 0.746085
## factor(msa)7160 27.636760 18.611153 1.485 0.138333
## factor(msa)7200 -9.518157 31.082845 -0.306 0.759595
## factor(msa)7240 -0.269025 22.383567 -0.012 0.990416
## factor(msa)7320 18.031659 10.586529 1.703 0.089286 .
## factor(msa)7360 14.678316 12.046036 1.219 0.223736
## factor(msa)7400 30.457513 22.495125 1.354 0.176504
## factor(msa)7485 0.164125 31.267953 0.005 0.995815
## factor(msa)7490 -0.237313 31.263089 -0.008 0.993947
## factor(msa)7520 10.517674 22.330859 0.471 0.637899
## factor(msa)7560 -9.896986 31.076444 -0.318 0.750291
## factor(msa)7600 45.914967 15.314750 2.998 0.002884 **
## factor(msa)7620 -9.588592 31.072297 -0.309 0.757792
## factor(msa)7680 11.281230 22.416778 0.503 0.615063
## factor(msa)7760 -10.159987 22.378505 -0.454 0.650067
## factor(msa)7800 29.109676 18.546165 1.570 0.117294
## factor(msa)7840 30.399450 31.263984 0.972 0.331458
## factor(msa)8040 22.667746 18.563769 1.221 0.222768
## factor(msa)8120 -11.523655 31.148165 -0.370 0.711604
## factor(msa)8160 5.170458 16.283003 0.318 0.751000

```

```
## factor(msa)8240    5.131228  31.249264   0.164 0.869653
## factor(msa)8280   -3.485327  15.122396  -0.230 0.817839
## factor(msa)8400   41.888537  22.373838   1.872 0.061898 .
## factor(msa)8560   63.057113  18.549710   3.399 0.000742 ***
## factor(msa)8640   -9.928865  31.102819  -0.319 0.749719
## factor(msa)8720   -9.486035  31.075014  -0.305 0.760323
## factor(msa)8735   -4.976588  22.372581  -0.222 0.824082
## factor(msa)8760   89.996568  31.105491   2.893 0.004018 **
## factor(msa)8800   -9.486035  31.075014  -0.305 0.760323
## factor(msa)8840   37.943572  10.538579   3.600 0.000357 ***
## factor(msa)8960   -5.109816  22.713672  -0.225 0.822119
## factor(msa)9000   87.099495  31.245657   2.788 0.005561 **
## factor(msa)9040   20.380365  31.250649   0.652 0.514669
## factor(msa)9140   -9.519582  31.073197  -0.306 0.759488
## factor(msa)9240   -9.283290  22.447013  -0.414 0.679412
## factor(msa)9320   45.054594  31.080022   1.450 0.147935
## factor(msa)9999    1.571479   6.856911   0.229 0.818843
## evercmptr        -10.321833   4.200104  -2.458 0.014408 *
## logtaxis         -5.712892   3.854468  -1.482 0.139079
## logtaxisq         1.198826   0.581455   2.062 0.039867 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 30.56 on 405 degrees of freedom
## Multiple R-squared:  0.3852, Adjusted R-squared:  0.1272
## F-statistic: 1.493 on 170 and 405 DF, p-value: 0.0007044
```

```
coeftest(treatment_msa, vcov = vcovHC(treatment_msa, "HC1"))
```

```
##
## t test of coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  16.284401  10.337202  1.5753 0.1159634
## factor(msa)80    0.424847   6.434626  0.0660 0.9473904
## factor(msa)160   19.344142  23.635008  0.8185 0.4135799
## factor(msa)200    0.126669   6.658892  0.0190 0.9848324
## factor(msa)220    0.465207   5.864267  0.0793 0.9368101
## factor(msa)240  -10.079845   4.962519 -2.0312 0.0428878 *
## factor(msa)280    2.911408   4.881686  0.5964 0.5512453
## factor(msa)460    0.464176   5.885436  0.0789 0.9371761
## factor(msa)520   19.846804  17.334079  1.1450 0.2529024
## factor(msa)560    4.218403  10.896018  0.3872 0.6988479
## factor(msa)600   24.096704  19.953883  1.2076 0.2278983
## factor(msa)640   48.179157  20.069083  2.4007 0.0168156 *
## factor(msa)720   12.998090  16.024219  0.8112 0.4177543
## factor(msa)730    0.600180   5.879546  0.1021 0.9187443
## factor(msa)760    3.196467   5.844891  0.5469 0.5847609
## factor(msa)870   -9.518157   4.808837 -1.9793 0.0484584 *
## factor(msa)875   -8.511729   5.906508 -1.4411 0.1503358
## factor(msa)1010 -11.126842   5.193664 -2.1424 0.0327579 *
## factor(msa)1120  -3.715275   7.635044 -0.4866 0.6267993
## factor(msa)1260  21.878263   6.283659  3.4818 0.0005523 ***
## factor(msa)1280   9.283253  11.944091  0.7772 0.4374795
```

```

## factor(msa)1305 -10.236697 5.000828 -2.0470 0.0413029 *
## factor(msa)1400 -10.195163 5.301689 -1.9230 0.0551806 .
## factor(msa)1440 14.917001 6.073399 2.4561 0.0144636 *
## factor(msa)1480 -10.884517 5.144262 -2.1159 0.0349668 *
## factor(msa)1520 6.568331 11.721970 0.5603 0.5755549
## factor(msa)1540 31.133557 5.492042 5.6688 2.735e-08 ***
## factor(msa)1580 -10.195163 5.301689 -1.9230 0.0551806 .
## factor(msa)1600 26.294442 14.695336 1.7893 0.0743128 .
## factor(msa)1640 36.387408 15.129441 2.4051 0.0166170 *
## factor(msa)1760 15.083255 6.084849 2.4788 0.0135880 *
## factor(msa)1840 19.093520 24.925356 0.7660 0.4441059
## factor(msa)1920 66.763254 20.696916 3.2258 0.0013584 **
## factor(msa)1960 -12.900505 7.481798 -1.7243 0.0854254 .
## factor(msa)2000 -0.866381 5.878276 -0.1474 0.8829000
## factor(msa)2080 55.007789 9.911963 5.5496 5.183e-08 ***
## factor(msa)2160 16.201016 11.845497 1.3677 0.1721663
## factor(msa)2240 -11.210043 5.923600 -1.8924 0.0591456 .
## factor(msa)2320 67.009437 5.544602 12.0855 < 2.2e-16 ***
## factor(msa)2400 -6.492459 6.120147 -1.0608 0.2893976
## factor(msa)2520 -5.155626 6.639058 -0.7766 0.4378718
## factor(msa)2560 0.476998 5.643716 0.0845 0.9326861
## factor(msa)2680 13.727898 9.503405 1.4445 0.1493648
## factor(msa)2720 37.717821 5.413899 6.9668 1.319e-11 ***
## factor(msa)2760 -9.707061 4.959581 -1.9572 0.0510065 .
## factor(msa)2840 -11.455069 6.283659 -1.8230 0.0690415 .
## factor(msa)2900 -0.075184 5.842344 -0.0129 0.9897388
## factor(msa)2920 -10.195163 5.301689 -1.9230 0.0551806 .
## factor(msa)2975 -9.605237 4.833703 -1.9871 0.0475803 *
## factor(msa)2995 -9.553375 4.812446 -1.9851 0.0478031 *
## factor(msa)3000 56.112706 5.519158 10.1669 < 2.2e-16 ***
## factor(msa)3060 0.241988 5.850593 0.0414 0.9670282
## factor(msa)3080 -10.195163 5.301689 -1.9230 0.0551806 .
## factor(msa)3120 31.981947 29.839422 1.0718 0.2844473
## factor(msa)3150 -1.255259 4.881686 -0.2571 0.7972041
## factor(msa)3160 22.367774 5.160979 4.3340 1.849e-05 ***
## factor(msa)3280 -5.340784 6.132262 -0.8709 0.3843072
## factor(msa)3320 12.696324 14.328725 0.8861 0.3761030
## factor(msa)3360 39.878405 41.891436 0.9519 0.3416919
## factor(msa)3400 -10.640533 5.092562 -2.0894 0.0372929 *
## factor(msa)3480 -8.929173 5.603607 -1.5935 0.1118352
## factor(msa)3520 -9.707061 4.959581 -1.9572 0.0510065 .
## factor(msa)3560 -9.490862 4.802450 -1.9763 0.0488041 *
## factor(msa)3600 -4.594232 5.820673 -0.7893 0.4304011
## factor(msa)3605 89.763303 5.000828 17.9497 < 2.2e-16 ***
## factor(msa)3610 -12.995731 7.595511 -1.7110 0.0878514 .
## factor(msa)3620 -10.195163 5.301689 -1.9230 0.0551806 .
## factor(msa)3640 -7.587954 5.237825 -1.4487 0.1481996
## factor(msa)3660 -10.195163 5.301689 -1.9230 0.0551806 .
## factor(msa)3760 55.386396 27.343917 2.0255 0.0434665 *
## factor(msa)3800 -9.896986 5.088768 -1.9449 0.0524826 .
## factor(msa)3810 -9.588592 4.881686 -1.9642 0.0501908 .
## factor(msa)3850 -9.605237 4.833703 -1.9871 0.0475803 *
## factor(msa)3980 -9.896986 5.088768 -1.9449 0.0524826 .
## factor(msa)4000 -14.027673 5.739493 -2.4441 0.0149487 *

```

```

## factor(msa)4040 0.005291 5.843655 0.0009 0.9992780
## factor(msa)4080 -9.605237 4.833703 -1.9871 0.0475803 *
## factor(msa)4100 -9.612609 4.828556 -1.9908 0.0471763 *
## factor(msa)4120 -13.670117 6.064844 -2.2540 0.0247305 *
## factor(msa)4200 23.854716 4.802626 4.9670 1.004e-06 ***
## factor(msa)4240 7.148509 4.808837 1.4865 0.1379155
## factor(msa)4280 -4.243591 6.018853 -0.7050 0.4811845
## factor(msa)4320 -10.236697 5.000828 -2.0470 0.0413029 *
## factor(msa)4400 -4.891515 5.400276 -0.9058 0.3655856
## factor(msa)4480 1.767368 7.849326 0.2252 0.8219670
## factor(msa)4520 -6.792289 7.054818 -0.9628 0.3362287
## factor(msa)4560 -9.490862 4.802450 -1.9763 0.0488041 *
## factor(msa)4600 -10.157708 4.981758 -2.0390 0.0421008 *
## factor(msa)4720 -6.967245 6.593131 -1.0567 0.2912588
## factor(msa)4900 -11.288861 4.942743 -2.2839 0.0228927 *
## factor(msa)4920 29.627492 5.974091 4.9593 1.043e-06 ***
## factor(msa)5015 33.530276 21.765991 1.5405 0.1242221
## factor(msa)5080 -0.486916 6.956531 -0.0700 0.9442329
## factor(msa)5120 -9.958299 4.845167 -2.0553 0.0404903 *
## factor(msa)5140 0.614772 6.282833 0.0978 0.9221002
## factor(msa)5160 8.948322 5.170208 1.7307 0.0842586 .
## factor(msa)5190 21.282168 26.248741 0.8108 0.4179635
## factor(msa)5200 0.465207 5.864267 0.0793 0.9368101
## factor(msa)5330 -10.668569 5.658614 -1.8854 0.0600957 .
## factor(msa)5345 -9.490862 4.802450 -1.9763 0.0488041 *
## factor(msa)5380 -3.191774 11.010004 -0.2899 0.7720429
## factor(msa)5400 -1.270511 8.433171 -0.1507 0.8803218
## factor(msa)5480 1.196967 10.480846 0.1142 0.9091317
## factor(msa)5560 31.073725 9.574814 3.2454 0.0012705 **
## factor(msa)5600 41.489552 16.184752 2.5635 0.0107225 *
## factor(msa)5640 4.424069 14.691778 0.3011 0.7634733
## factor(msa)5660 -9.505222 4.808037 -1.9769 0.0487257 *
## factor(msa)5720 -10.218472 5.664065 -1.8041 0.0719603 .
## factor(msa)5775 23.789338 15.380854 1.5467 0.1227202
## factor(msa)5800 -9.112704 4.886519 -1.8649 0.0629232 .
## factor(msa)5920 -5.077944 6.102336 -0.8321 0.4058252
## factor(msa)5945 -6.089332 6.227197 -0.9779 0.3287269
## factor(msa)6080 -1.238671 8.997216 -0.1377 0.8905675
## factor(msa)6120 -5.841000 5.993650 -0.9745 0.3303745
## factor(msa)6160 4.781868 9.514610 0.5026 0.6155316
## factor(msa)6200 15.254947 15.893898 0.9598 0.3377290
## factor(msa)6280 -11.530757 5.656207 -2.0386 0.0421387 *
## factor(msa)6320 -9.557124 4.819727 -1.9829 0.0480517 *
## factor(msa)6340 0.424847 6.434626 0.0660 0.9473904
## factor(msa)6400 -9.525769 4.802371 -1.9836 0.0479803 *
## factor(msa)6440 31.198837 12.354445 2.5253 0.0119402 *
## factor(msa)6480 10.197977 17.519913 0.5821 0.5608372
## factor(msa)6640 44.260565 16.241584 2.7251 0.0067061 **
## factor(msa)6680 -10.984777 5.044597 -2.1775 0.0300174 *
## factor(msa)6760 -3.520278 7.279489 -0.4836 0.6289392
## factor(msa)6780 6.686464 11.174536 0.5984 0.5499301
## factor(msa)6800 -7.948482 5.898086 -1.3476 0.1785284
## factor(msa)6840 89.460909 5.095879 17.5555 < 2.2e-16 ***
## factor(msa)6920 40.134531 19.615233 2.0461 0.0413928 *

```

```

## factor(msa)6960 -9.518157 4.808837 -1.9793 0.0484584 *
## factor(msa)7040 -3.903501 5.470028 -0.7136 0.4758755
## factor(msa)7080 -10.079845 4.962519 -2.0312 0.0428878 *
## factor(msa)7160 27.636760 32.206480 0.8581 0.3913383
## factor(msa)7200 -9.518157 4.808837 -1.9793 0.0484584 *
## factor(msa)7240 -0.269025 5.439177 -0.0495 0.9605766
## factor(msa)7320 18.031659 13.650660 1.3209 0.1872683
## factor(msa)7360 14.678316 12.857676 1.1416 0.2542951
## factor(msa)7400 30.457513 19.595548 1.5543 0.1208921
## factor(msa)7485 0.164125 5.847697 0.0281 0.9776229
## factor(msa)7490 -0.237313 5.840842 -0.0406 0.9676110
## factor(msa)7520 10.517674 9.703538 1.0839 0.2790535
## factor(msa)7560 -9.896986 5.088768 -1.9449 0.0524826 .
## factor(msa)7600 45.914967 16.885457 2.7192 0.0068254 **
## factor(msa)7620 -9.588592 4.881686 -1.9642 0.0501908 .
## factor(msa)7680 11.281230 6.821190 1.6539 0.0989327 .
## factor(msa)7760 -10.159987 4.981177 -2.0397 0.0420310 *
## factor(msa)7800 29.109676 20.247200 1.4377 0.1512875
## factor(msa)7840 30.399450 5.841422 5.2041 3.104e-07 ***
## factor(msa)8040 22.667746 32.361375 0.7005 0.4840443
## factor(msa)8120 -11.523654 5.271439 -2.1861 0.0293835 *
## factor(msa)8160 5.170458 16.120814 0.3207 0.7485790
## factor(msa)8240 5.131228 5.942534 0.8635 0.3883875
## factor(msa)8280 -3.485327 7.351024 -0.4741 0.6356641
## factor(msa)8400 41.888537 43.605830 0.9606 0.3373175
## factor(msa)8560 63.057113 24.707831 2.5521 0.0110735 *
## factor(msa)8640 -9.928865 4.923817 -2.0165 0.0444076 *
## factor(msa)8720 -9.486035 4.812411 -1.9712 0.0493859 *
## factor(msa)8735 -4.976588 6.690435 -0.7438 0.4574069
## factor(msa)8760 89.996568 4.943174 18.2062 < 2.2e-16 ***
## factor(msa)8800 -9.486035 4.812411 -1.9712 0.0493859 *
## factor(msa)8840 37.943572 13.979036 2.7143 0.0069250 **
## factor(msa)8960 -5.109816 6.972867 -0.7328 0.4640956
## factor(msa)9000 87.099495 7.481798 11.6415 < 2.2e-16 ***
## factor(msa)9040 20.380365 5.877113 3.4678 0.0005811 ***
## factor(msa)9140 -9.519582 4.836398 -1.9683 0.0497128 *
## factor(msa)9240 -9.283290 8.592978 -1.0803 0.2806362
## factor(msa)9320 45.054594 4.802450 9.3816 < 2.2e-16 ***
## factor(msa)9999 1.571479 6.829477 0.2301 0.8181284
## evercmptr -10.321833 4.259647 -2.4232 0.0158234 *
## logtaxis -5.712892 5.478506 -1.0428 0.2976709
## logtaxisq 1.198826 0.767675 1.5616 0.1191556
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Q3)

- a) The coefficient for evercmptr is about -8.776, which implies that when computer is adopted, the driver ownership rate decreases by about 8.776%.
- b) This might be because the within-cluster correlation is 0. In other words, geographical variation does not help explain driver ownership.
- c) After adding in the MSA fixed effects, the model's R^2 becomes larger, and the model's AIC becomes smaller, suggesting that adding the MSA fixed effects did improve the model. Comparing the results

based on robust standard errors, we can also see that, after adding in the MSA fixed effects, the coefficients for logtaxi and logtaxisq are no longer significant at 0.05 level, while the coefficient for evercmptr, although still significant at 0.05 level, becomes less significant than before. This suggests that firm size (logtaxi and logtaxisq) might be explained by local market variation (MSA).

- d) From the results, we can see that the higher the computer adoption rate is, the lower the driver ownership rate is. In other words, the taxis are more likely to be owned by the contracting firm rather than the drivers themselves when computers are adopted more and more. This might be because computers allow companies to better monitor how drivers drive their cars.
- e) Firm size might actually determine whether the firm adopts technologies or not. So adoption might actually be endogenous.

```
## Q4

# 2SLS
# 1st stage: regress endogenous regressor on instruments
first_stage <- lm(evercmptr ~ northeast + midwestdum + southdum + worktransit, data = taxi_no_nas)
summary(first_stage)

##
## Call:
## lm(formula = evercmptr ~ northeast + midwestdum + southdum +
##     worktransit, data = taxi_no_nas)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.7895 -0.3236 -0.2696  0.5574  0.9120
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.320852   0.034698   9.247  < 2e-16 ***
## northeast   -0.245484   0.048966  -5.013 7.15e-07 ***
## midwestdum  -0.051971   0.051399  -1.011   0.312
## southdum    -0.046390   0.055326  -0.838   0.402
## worktransit  0.013990   0.002424   5.772 1.28e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4438 on 571 degrees of freedom
## Multiple R-squared:  0.0771, Adjusted R-squared:  0.07063
## F-statistic: 11.92 on 4 and 571 DF,  p-value: 2.595e-09

# 2nd stage: regress dependent variable on fitted values from 1st stage along with other independent variables
second_stage <- lm(downrt ~ first_stage$fitted.values + logtaxi + logtaxisq, data = taxi_no_nas)
summary(second_stage)

##
## Call:
## lm(formula = downrt ~ first_stage$fitted.values + logtaxi +
##     logtaxisq, data = taxi_no_nas)
##
## Residuals:
```

```
##      Min      1Q Median      3Q      Max
## -53.88 -18.62 -13.36   8.28  91.93
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      12.4980      6.1282   2.039  0.04187 *
## first_stage$fitted.values 45.4410     11.0107   4.127 4.22e-05 ***
## logtaxis         -7.4006      3.3042  -2.240  0.02549 *
## logtaxisq         1.4263      0.4938   2.888  0.00402 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 31.63 on 572 degrees of freedom
## Multiple R-squared:  0.07026,    Adjusted R-squared:  0.06538
## F-statistic: 14.41 on 3 and 572 DF,  p-value: 4.634e-09
```

Q4)

- a) The F-statistics for the first stage model is 11.92, with a p-value of 2.595e-09, suggesting that the model is significantly different from a null model. The coefficient for northeast is about -0.245, with a p-value of 7.15e-07, indicating that northeast has a higher computer adoption rate. The coefficient for worktransit is about 0.014, with a p-value of 1.28e-08. The R^2 suggests that 7.71% of the variance of evercmptr is explained by the model.
- b) The second stage least square has a F of 14.41 and a R^2 of 0.07026, which are both larger than their comparables in the OLS model, suggesting that the 2SLS performs better than the OLS. However, the coefficient for evercmptr_hat is 45.44, while it is -8.776 in the OLS model, indicating that the instruments we selected actually are important in our model. In other words, regions and worktransit is correlated with driver ownership only because it is correlated with computer adoption, so after we factor these instruments out, an increase in computer adoption actually leads to a increase in predicted driver ownership.
- c) From the results, we can see that the higher the computer adoption rate is, the lower the driver ownership rate is, if we don't account for endogeneity. But if we do account for endogeneity, the effect reverses. In other words, the taxis are more likely to be owned by the contracting firm rather the drivers themselves when computers are less adopted, after we take endogeneity into account. An ideal experiment would be randomly assigning taxis to treatment and control conditions that differ in the level of computer adoption and observe how driver ownership change over time. It could be difficult to randomly assign computer technology to taxis, so the best we could do is perhaps use difference-in-difference to factor out taxi-specific and time-specific effects.