Logit Exercises

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1. Specify a logit demand model model, in which you include observable product characteristics.

```
setwd('~/Box Sync/abarciausksas/15D011 Economics for the Era of Big Data/Christian Michel/PS3/')
library(foreign)
# Load data
data <- read.dta('cars.dta')</pre>
households <- unique(data$pop) / 3
outside_market_share <- (households - sum(data[,'qu'])) / households
data$inside_market_share <- data$qu/households
# hp: horsepower
# li: fuel consumption in liters per 100 km
# wi, he: width and height in cm
# cy: displacement (in cc)
# cla: factor(segment)
data$demand <- log(data$inside_market_share) - log(outside_market_share)</pre>
data$brd <- factor(data$brd)</pre>
data$cla <- factor(data$cla)</pre>
model.1 <- lm(demand ~ hp + li + wi + he + cy + cla + princ, data = data)
summary(model.1)
##
## Call:
## lm(formula = demand ~ hp + li + wi + he + cy + cla + princ, data = data)
## Residuals:
##
      Min
               1Q Median
                                3Q
                                       Max
## -3.7849 -0.8154 0.1880 0.9111 2.6624
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.304e+01 3.485e+00 -6.612 1.85e-10 ***
              -6.744e-02 1.046e-02 -6.446 4.85e-10 ***
## hp
              3.085e-02 7.756e-02 0.398 0.69111
## li
## wi
              1.289e-01 1.691e-02 7.627 3.54e-13 ***
              -1.071e-02 1.984e-02 -0.540 0.58959
## he
## cy
              -1.235e-03 5.899e-04 -2.093 0.03724 *
## cla2
              5.717e-01 2.791e-01 2.048 0.04143 *
```

```
## cla3
               4.365e-01 3.388e-01
                                     1.288 0.19865
## cla4
               4.000e-01 4.865e-01
                                     0.822 0.41165
## cla5
               6.437e-01 6.317e-01
                                     1.019 0.30905
               1.946e+00 6.487e-01
                                     3.000 0.00294 **
## princ
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 1.25 on 287 degrees of freedom
## Multiple R-squared: 0.3254, Adjusted R-squared: 0.3019
## F-statistic: 13.84 on 10 and 287 DF, p-value: < 2.2e-16
```

2. Discuss endogeneity problems and how you are going to treat them.

Endogeneity problems arise when unobservable characteristics are correlated with the regressors. In this dataset, the effect of price may be correlated with unobservable characteristics like style. Therefore, any estimate of the effect of price may be biased by these unobservable characteristics.

We are going to treat them by using an instrumental variable which is correlated with price but not with unobservables. We expect the instrumental variable to be correlated with the demand but only through price. We have selected to sum the characteristics of products in other segments as the instrumental variable.

3. Estimate the demand equation of the logit model with and without instrumenting for price.

We estimated the demand equation of the logit model without instrumenting for price in part 1.

```
# Want to regress price on our instrumental variables which will be defined as the sum of the character
calculate_iv <- function(row, characteristic) {</pre>
  # other_products not in our segment
  segment <- row['cla']</pre>
  other_products <- subset(data, cla != segment)</pre>
  sum(other_products[,characteristic])
}
data$iv.hp <- apply(data, 1, function(row) calculate_iv(row, 'hp'))</pre>
data$iv.li <- apply(data, 1, function(row) calculate_iv(row, 'li'))</pre>
data$iv.wi <- apply(data, 1, function(row) calculate_iv(row, 'wi'))</pre>
data$iv.he <- apply(data, 1, function(row) calculate_iv(row, 'he'))</pre>
data$iv.cy <- apply(data, 1, function(row) calculate_iv(row, 'cy'))</pre>
price_model <- lm(princ ~ iv.hp + iv.li + iv.wi + iv.he + iv.cy, data = data)</pre>
data$priceiv <- fitted.values(price_model)</pre>
# Instrumented price model
model.2 \leftarrow lm(demand \sim hp + li + wi + he + cy + priceiv, data = data)
summary(model.2)
##
## Call:
## lm(formula = demand ~ hp + li + wi + he + cy + priceiv, data = data)
##
## Residuals:
```

```
##
               10 Median
                               3Q
                                      Max
## -3.6667 -0.7524 0.1534 1.0122
                                  2.3837
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.641e+01 3.373e+00 -7.829 9.22e-14 ***
## hp
              -5.549e-02 9.839e-03
                                     -5.639 4.04e-08 ***
## li
               1.037e-02 7.790e-02
                                      0.133
                                              0.8941
## wi
               1.434e-01
                          1.521e-02
                                      9.425
                                             < 2e-16 ***
## he
              -4.888e-03 1.959e-02
                                     -0.250
                                              0.8031
## cy
              -9.401e-04 5.778e-04
                                     -1.627
                                              0.1048
                                              0.0796 .
## priceiv
               1.020e+00 5.796e-01
                                      1.759
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.268 on 291 degrees of freedom
## Multiple R-squared: 0.2971, Adjusted R-squared: 0.2826
## F-statistic: 20.5 on 6 and 291 DF, p-value: < 2.2e-16
```

4. Price elasticities

```
alpha <- coefficients(model.2)['priceiv'][[1]]</pre>
# Own price elasticity is the elasticity of demand for project j given a change in price j
data$elasticity.own <- -alpha*(1-data$inside_market_share)*(data$princ)</pre>
# Cross price elasticity is the elasticity of demand for project x given a change in price j
# We can calculate the cross price elasticity for every product
data$elasticity.cross <- alpha*(data$inside_market_share)*data$princ</pre>
summary(data$elasticity.own)
      Min. 1st Qu. Median
                               Mean 3rd Qu.
## -2.2050 -0.8259 -0.5818 -0.6774 -0.4131 -0.2520
summary(data$elasticity.cross)
##
        Min.
               1st Qu.
                           Median
                                       Mean
                                               3rd Qu.
                                                            Max.
## 4.653e-06 1.070e-04 3.276e-04 7.071e-04 9.919e-04 4.372e-03
## Marginal cost
data$marginal_cost <- data$princ * (1 + 1/data$elasticity.own)</pre>
summary(data$marginal_cost)
##
                               Mean 3rd Qu.
      Min. 1st Qu. Median
                                                Max.
## -0.7339 -0.5763 -0.4109 -0.3168 -0.1711
```

General patterns:

• Cross-price elasticity is positive but very near zero, with values between near 4.653e-06 and 4.372e-03. The interpretation is that demand for product j when the price in product k increases is increasing but approximately zero. This interpretation could be restated as the change in price of any other product has negligible effect on the demand for product k.

- Own-price elasticity is negative with values between -2.205 and -0.25. The interestation is that demand for product j when the price in product j increses is decreasing.
- Marginal cost is positive between -0.734 and 1.18.