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18 September 2014

Online at <https://mpra.ub.uni-muenchen.de/59117/>

MPRA Paper No. 59117, posted 07 Oct 2014 14:04 UTC

Demand Model Simulation in R with Endogenous Prices and Unobservable Quality

Daniel Toro Gonzalez*

Abstract

The aim of the present routine is to simulate a demand equation with endogenous prices and unobservable product quality and to retrieve the original parameters using the Control Function (CF) approach. The CF approach is a very useful and simple method to obtain unbiased estimates. The present R code helps to understand the underlying structure of the endogeneity problem in demand estimations. Results support the important bias correction of the CF approach.

Key Words: R, Demand, Endogeneity, Simulation, Endogenous Prices

JEL: C13, C15, D49, E27

Introduction

The exercise consists in the repeated estimation of three models by OLS, the original model designed by simulation of the variables, non-observable quality model which omits the quality variable from the specification, and the Control Function (CF) model which is estimated in two stages.

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Routine

The software can be downloaded from the *R Project for Statistical Computing* at the web site www.r-project.org.

```
# SIMULATE A Demand model with ENDOGENOUS prices1
# Daniel Toro Gonzalez 2014
# Universidad Tecnológica de Bolívar

remove()           # Remove objects from the workspace
rm(list=ls())      # Clearing memory space

#####
# NECESSARY PACKAGES TO RUN THIS PROGRAM#
#####

#install.packages("AER") #In case is not installed remove
                          # the first # sign infront of the line

#install.packages("systemfit") #In case is not installed remove
                                #the first # sign infront of the line
library(AER)                 #Load library AER
library(systemfit)           #Load library Systemfit

#####

# The data generating model, unobservable by the researcher is:
#  $DD = B_0 + B_1 * P + B_2 * I + B_3 * X + e_1$  with  $e_1 \sim N(0,1)$ 
# Where DD are the quantities demanded, P are the prices, I is the income
# and X quality of the product.

#####
# MODEL SETUP#
#####

n <- 50                #Sample size (Change at will)
S <- 1000              #Number of samples (Change at will)

# Parameter Values (Change at will)
B0 <- 100              #Intercept of Demand
B1 <- -2               #Price coefficient
B2 <- +5               #Income coefficient
B3 <- +3               #Quality coefficient

parameters <- c(B0, B1, B2, B3) # Save the original
                                # population parameters for printing purposes

#Creates empty matrices to store the parameters in each trial
results0 <- matrix(nrow=S,ncol=4)
results1 <- matrix(nrow=S,ncol=3) # There is no coefficient for X
results2 <- matrix(nrow=S,ncol=4)

# The following Loop generates S number of repetitions (or samples)
```

¹ Created by Pretty R at inside-R.org

Toro (2014)

```
for (i in 1:S){

  set.seed(i)
  # Errors
  e1 <- rnorm(n, mean=0, sd=1)      #Error e1~N(0,k) k=1, this term is
included in the demand equation
  e2 <- rnorm(n, mean=0, sd=1)      #Error e2~N(0,k) k=1, this term will be
included in the endogenous price equation

  # Quality (Exogenous variable, non observable)
  X <- rnorm(n, mean=10, sd=1)      #Quality X~N(10,k) k=1

  # Price of Labor (Exogenous variable, observable)
  W <- rnorm(n, mean=18, sd=4)      #Minimum wage by hour W~N(18,k) k=1

  #Average Exogenous Income US$5000
  I <- rnorm(n, mean=5000, sd=1000) #Income I~N(5000,500)

  # GENERATING ENDOGENOUS PRICES
  # Endogenous Prices depend on labor cost (W), Quality (X) and a random
term e2~N(0,k) k=1
  # The values of the parameters (10, 0.5, 2) can be modified at will
  P <- 10+0.5*W+2*X+e2

  # Generating quantity values for the demand function
  # using the unobservable population parameters (B0, B1, B2, B3)
  # and the simulated values of the variables (P, I, X, e1)

  Q <- B0+B1*P+B2*I+B3*X+e1 # Demand Curve

  #####
  #          BASE ESTIMATION          #
  #####

  # Reg0 is the estimation of the demand model
  # as if quality is an observable variable
  # this model replicates closely the real data generating process
  reg0 <- lm(Q ~ P + I + X)
  results0[i,] <- coef(reg0)

  #####
  #          ENDOGENOUS PRICE ESTIMATION          #
  #####

  # Reg1 is the estimation omiting Quality
  # which is unobservable but highly correlated with the prices
  # this model yields bias estimators for the price parameter
  reg1 <- lm(Q ~ P + I)
  results1[i,] <- coef(reg1)
```

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```
#####  
#      CONTROL FUNCTION      #  
#####  
  
# Stage No.1  
# Assume W is observable so we can estimate  $P = D0+D1W+u2$   
#  $P=10+0.5*W+2*X+e2$   
  
# ESTRUCTURAL PRICE EQUATION #  
  
regP <- lm(P ~ W)  
CF <- resid(regP) # The Control Function Variable is  
# the error term of the Estructural Price Equation  
  
# DEMAND EQUATION WITH CONTROL #  
# Stage No.2  
# We estimate the demand model using CF instead of X,  
# hence:  $Q = B0+B1*P+B2*I+B3*CF$   
reg2 <- lm(Q ~ P + I + CF)  
results2[i,] <- coef(reg2)  
  
}  
  
#####  
#      PRINT RESULTS      #  
#####  
  
colnames(results0) <- c("INTERC", "PRICE", "INCOME", "QUALITY")  
colnames(results1) <- c("INTERC", "PRICE", "INCOME")  
colnames(results2) <- c("INTERC", "PRICE", "INCOME", "QUALITY")  
  
summary(results0) # ORIGINAL MODEL  
summary(results1) # UNOBSERVABLE QUALITY  
# Bias in price coefficient (%)  
(mean(results1[,2]) - mean(results0[,2])) / mean(results0[,2]) * 100  
summary(results2) # CONTROL FUNCTION  
# Bias in price coefficient (%)  
(mean(results2[,2]) - mean(results0[,2])) / mean(results0[,2]) * 100
```

Results

The first set of coefficients corresponds to the original model, including all the variables price, income and quality, to explain the demand. For each of the one thousand samples, the set of three parameters is estimated and the summary statistics of the vector of parameters is presented in the next table. It is clear that since all the relevant variables of the model are included, the parameter values on average are very close to the original parameter values ($B0=100$, $B1=-2$, $B2=5$, $B3=3$).

```
> summary(results0) # ORIGINAL MODEL
      INTERC      PRICE      INCOME      QUALITY
Min.   : 91.82  Min.   :-2.215  Min.   :5    Min.   :2.231
1st Qu.: 98.49  1st Qu.:-2.042  1st Qu.:5    1st Qu.:2.873
Median :100.02  Median :-2.001  Median :5    Median :3.005
Mean   : 99.92  Mean   :-1.999  Mean   :5    Mean   :3.003
3rd Qu.:101.30  3rd Qu.:-1.957  3rd Qu.:5    3rd Qu.:3.135
Max.   :106.72  Max.   :-1.776  Max.   :5    Max.   :3.681
```

The second set of coefficients corresponds to the omitted quality model. In this case the variables explaining the demanded quantities are price and income. As in the previous case, for each of the one thousand samples the set of two parameters is estimated and the summary statistics of the vector of parameters is presented in the next table. In this case since not all the relevant variables of the model are included, the parameter values, specifically the price coefficient is biased on average. Compared to the original parameter of -2, the observed parameter is biased in -33% (-1.33).

```
> summary(results1) # UNOBSERVABLE QUALITY
      INTERC      PRICE      INCOME
Min.   : 88.78  Min.   :-1.8160  Min.   :4.999
1st Qu.:100.83  1st Qu.:-1.4111  1st Qu.:5.000
Median :104.04  Median :-1.3296  Median :5.000
Mean   :104.09  Mean   :-1.3334  Mean   :5.000
3rd Qu.:107.39  3rd Qu.:-1.2520  3rd Qu.:5.000
Max.   :120.02  Max.   :-0.9841  Max.   :5.001
```

Finally, when the control function method is implemented by using the errors of the structural price equation to replace the variable Quality, the results show that the price coefficient is no longer biased.

```
> summary(results2) # CONTROL FUNCTION
      INTERC      PRICE      INCOME      QUALITY
Min.   :109.2  Min.   :-4.016  Min.   :4.999  Min.   :0.4947
1st Qu.:124.0  1st Qu.:-2.158  1st Qu.:5.000  1st Qu.:1.0356
Median :129.5  Median :-1.985  Median :5.000  Median :1.1964
Mean   :130.9  Mean   :-2.024  Mean   :5.000  Mean   :1.2237
3rd Qu.:136.4  3rd Qu.:-1.848  3rd Qu.:5.000  3rd Qu.:1.3803
Max.   :205.9  Max.   :-1.458  Max.   :5.001  Max.   :3.2646
```

Conclusion

The CF strategy allows the researcher to control for the unobservable factors and to correctly identify the price parameter.