

SCRIPT MOD4S1B: ENDOGENEITY AND TSLS

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LOAD AND PREPARE DATA

This example is based on Greene's consumption regression in Example 5.3 (5th edition). As shown in the lecture notes, in this case the correlation between \mathbf{X} and the error term arises due to endogeneity.

```
R> data<- read.table('c:/Klaus/AAEC5126/R/data/consumption.txt', sep="\t", header=FALSE)
R> #
R> #assign variable names
R> names(data)[1]<-"year"
R> names(data)[2]<-"quarter"
R> names(data)[3]<-"realgdp"
R> names(data)[4]<-"realcons"
R> names(data)[5]<-"realinv"
R> names(data)[6]<-"realgovt"
R> names(data)[7]<-"realdpi"
R> names(data)[8]<-"cpi"
R> names(data)[9]<-"m1"
R> names(data)[10]<-"tbill"
R> names(data)[11]<-"unemp"
R> names(data)[12]<-"pop"
R> names(data)[13]<-"infl"
R> names(data)[14]<-"realint"
R> #
R> save(data, file = "c:/Klaus/AAEC5126/R/data/consumption.rda")
R> attach(data)
```

Variable definitions:

```
% Contents of data
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%
% 1      Year = Date
% 2      Qtr = Quarter
% 3      Realgdp = Real GDP ($billion)
% 4      Realcons = Real consumption expenditures ($billion)
% 5      Realinvs = Real investment by private sector ($billion)
% 6      Realgovt = Real government expenditures ($billion)
% 7      Realdpi = Real disposable personal income ($billion)
% 8      CPI_U = Consumer price index
% 9      M1 = Nominal money stock
% 10     Tbilrate = Quarterly average of month end 90 day t bill rate
% 11     Unemp = Unemployment rate
% 12     pop = Population, million
% 13     Infl = Rate of inflation (first observation is missing and set to zero)
% 14     Realint = Ex post real interest rate = Tbilrate - Infl.
```

```
%      (First observation missing and set to zero)
```

SIMPLE OLS

```
R> # Define variables
R> n<-nrow(data)
R> y<-realcons[2:n]
R> ylag<-realcons[1:(n-1)]
R> dpi<-realdpi[2:n]
R> dpilag<-realdpi[1:(n-1)]
R> n<-length(y) #IMPORTANT - re-define n!
R> #
R> X<-cbind(rep(1,n),dpi)
R> k<-ncol(X)
R> #
R> bols<-solve((t(X)) %*% X) %*% (t(X) %*% y)# compute OLS estimator
R> e<-y-X%*%bols # Get residuals.
R> SSR<-(t(e)%*%e)#sum of squared residuals - should be minimized
R> s2<-(t(e)%*%e)/(n-k) #get the regression error (estimated variance of "eps").
R> s2ols<-s2 #for Hausman test below
R> Vb<-s2[1,1]*solve((t(X))%*%X) # get the estimated VCOV matrix of bols
R> se=sqrt(diag(Vb)) # get the standard errors for your coefficients;
R> tval=bols/se # get your t-values.
R> #
R> tt<-data.frame(col1=c("constant","dpi"),
                 col2=bols,
                 col3=se,
                 col4=tval)
R> colnames(tt)<-c("variable","estimate","s.e.,"t")
```

TABLE 1. OLS output

variable	estimate	s.e.	t
constant	-81.24659	14.42952	-5.63058
dpi	0.92188	0.00390	236.58551

TOLS

We will instrument dpi by lagged dpi and lagged consumption.

```
R> # Build instrument matrix
R> Z<-cbind(rep(1,n),dpilag,ylag)
R> Xhat<-Z %*% solve(t(Z) %*% Z) %*% t(Z) %*% X
R> k<-ncol(Xhat) #Don't forget to update k!
R> #
R> btols<-solve((t(Xhat)) %*% Xhat) %*% (t(Xhat) %*% y)# compute OLS estimator
R> e<-y-X%*%btols # careful - don't use Xhat here!
R> SSR<-(t(e)%*%e)#sum of squared residuals - should be minimized
R> s2<-(t(e)%*%e)/n #no need to correct for k in this case
```

```

R> Vb<-s2[1,1]*solve((t(Xhat))%*%Xhat) # get the estimated VCOV matrix of bols
R> se=sqrt(diag(Vb)) # get the standard errors for your coefficients;
R> tval=btsls/se # get your t-values.
R> #
R> tt<-data.frame(col1=c("constant","dpi"),
                  col2=btsls,
                  col3=se,
                  col4=tval)
R> colnames(tt)<-c("variable","estimate","s.e.,"t")

```

TABLE 2. TSLS output

variable	estimate	s.e.	t
constant	-81.95259	14.36040	-5.70685
dpi	0.92209	0.00388	237.77128

HAUSMAN TEST

```

R> d<-btsls-bols
R> W<-solve(t(Xhat) %*% Xhat)- solve(t(X) %*% X)
R> H<-(t(d) %*% pseudoinverse(W) %*% d)/s2ols[1,1] #Note use of OLS s2
R> J<-1
R> pval=1-pchisq(H,J)

```

The Hausman test statistic is 8.4814. The corresponding p-value is 0.0036.

WU TEST

```

R> # Step 1: regress dpi on Z and capture predicted values
R> dpihat<- Z %*% solve(t(Z) %*% Z) %*% t(Z) %*% dpi
R> #
R> # Step 2: add predicted values to original regression
R> X<-cbind(rep(1,n),dpi,dpihat)
R> k<-ncol(X)
R> bwu<-solve((t(X)) %*% X) %*% (t(X) %*% y)# compute OLS estimator
R> e<-y-X%*%bwu # Get residuals.
R> s2<-(t(e)%*%e)/(n-k) #get the regression error (estimated variance of "eps").
R> Vb<-s2[1,1]*solve((t(X))%*%X) # get the estimated VCOV matrix of bols
R> #
R> # Step 3: Perform F-test
R> Rmat<-matrix(c(0, 0, 1),nrow=1)
R> q<- 0
R> J<-nrow(Rmat)
R> b<-bwu
R> Fstat<-(1/J)* t(Rmat %*% b-q) %*% solve(Rmat%*%Vb%*%t(Rmat))%*%(Rmat%*%b-q)
R> pval<-1-pf(Fstat,J,n-k)

```

The Wu test statistic is 8.811. The corresponding p-value is 0.0034.

```
R> proc.time()-tic
  user  system elapsed
 0.15   0.06   0.25
```