

MODULE TWO, PART TWO: ENDOGENEITY, INSTRUMENTAL VARIABLES AND TWO-STAGE LEAST SQUARES IN ECONOMIC EDUCATION RESEARCH USING LIMDEP

Part Two of Module Two provides a cookbook-type demonstration of the steps required to use LIMDEP to address problems of endogeneity using a two-stage least squares, instrumental variable estimator. The Durbin, Hausman and Wu specification test for endogeneity is also demonstrated. Users of this model need to have completed Module One, Parts One and Two, and Module Two, Part One. That is, from Module One, users are assumed to know how to get data into LIMDEP, recode and create variables within LIMDEP, and run and interpret regression results. From Module Two, Part One, they are expected to have an understanding of the problem of and source of endogeneity and the basic idea behind an instrumental variable approach and the two-stage least squares method. The Becker and Johnston (1999) data set is used throughout this module for demonstration purposes only. Module Two, Parts Three and Four demonstrate in STATA and SAS what is done here in LIMDEP.

THE CASE

As described in Module Two, Part One, Becker and Johnston (1999) called attention to classroom effects that might influence multiple-choice and essay type test taking skills in economics in different ways. For example, if the student is in a classroom that emphasizes skills associated with multiple choice testing (*e.g.*, risk-taking behavior, question analyzing skills, memorization, and keen sense of judging between close alternatives), then the student can be expected to do better on multiple-choice questions. By the same token, if placed in a classroom that emphasizes the skills of essay test question answering (*e.g.*, organization, good sentence and paragraph construction, obfuscation when uncertain, logical argument, and good penmanship), then the student can be expected to do better on the essay component. Thus, Becker and Johnston attempted to control for the type of class of which the student is a member. Their measure of “teaching to the multiple-choice questions” is the mean score or mark on the multiple-choice questions for the school in which the i^{th} student took the 12th grade economics course. Similarly, the mean school mark or score on the essay questions is their measure of the i^{th} student’s exposure to essay question writing skills.

In equation form, the two equations that summarize the influence of the various covariates on multiple-choice and essay test questions are written as the follow structural equations:

$$M_i = \rho_{21} + \rho_{22}W_i + \rho_{23}\bar{M}_i + \sum_{j=4}^J \rho_{2j}X_{ij} + U_i^*$$

$$W_i = \rho_{31} + \rho_{32}M_i + \rho_{33}\bar{W}_i + \sum_{j=4}^J \rho_{3j}X_{ij} + V_i^*$$

M_i and W_i are the i^{th} student's respective scores on the multiple-choice test and essay test. \bar{M}_i and \bar{W}_i are the mean multiple-choice and essay test scores at the school where the i^{th} student took the 12th grade economics course. The X_{ij} variables are the other exogenous variables (such as gender, age, English a second language, etc.) used to explain the i^{th} student's multiple-choice and essay marks, where the ρ s are parameters to be estimated. The inclusion of the mean multiple-choice and mean essay test scores in their respective structural equations, and their exclusion from the other equation, enables both of the structural equations to be identified within the system.

As shown in Module Two, Part One, the least squares estimation of the ρ s involves bias because the error term U_i^* is related to W_i , in the first equation, and V_i^* is related to M_i , in second equation. Instruments for regressors W_i and M_i are needed. Because the reduced form equations express W_i and M_i solely in terms of exogenous variables, they can be used to generate the respective instruments:

$$M_i = \Gamma_{21} + \Gamma_{22} \bar{W}_i + \Gamma_{23} \bar{M}_i + \sum_{j=4}^J \Gamma_{2j} X_{ij} + U_i^{**}.$$

$$W_i = \Gamma_{31} + \Gamma_{32} \bar{M}_i + \Gamma_{33} \bar{W}_i + \sum_{j=4}^J \Gamma_{3j} X_{ij} + V_i^{**}.$$

The reduced form parameters (Γ s) are functions of the ρ s, and the reduced form error terms U^{**} and V^{**} are functions of U^* and V^* , which are not related to any of the regressors in the reduced form equations.

We could estimate the reduced form equations and get \hat{M}_i and \hat{W}_i . We could then substitute \hat{M}_i and \hat{W}_i into the structural equations as proxy regressors (instruments) for M_i and W_i . The least squares regression of M_i on \hat{W}_i , \bar{M}_i and the X s and a least squares regression of W_i on \hat{M}_i , \bar{W}_i and the X s would yield consistent estimates of the respective ρ s, but the standard errors would be incorrect. LIMDEP automatically does all the required estimations with the two-stage, least squares command:

2SLS; LHS= ; RHS= ; INST= \$

TWO-STAGE, LEAST SQUARES IN LIMDEP

The Becker and Johnston (1999) data are in the file named "Bill.CSV." Before reading these data into LIMDEP, however, the "Project Settings" must be increased from 200000 cells (222

rows and 900 columns) to accommodate the 4,178 observations. This can be done with a project setting of 4000000 cells (4444 rows and 900 columns), following the procedures described in Module One, Part Two. After increasing the project setting, file Bill.CSV can be read into LIMDEP with the following read command (the file may be located anywhere on your hard drive but here it is located on the e drive):

```
READ; NREC=4178; NVAR=44; FILE=e:\bill.csv; Names=
student,school,size,other,birthday,sex,eslflag,adultst,
mc1,mc2,mc3,mc4,mc5,mc6,mc7,mc8,mc9,mc10,mc11,mc12,mc13,
mc14,mc15,mc16,mc17,mc18,mc19,mc20,totalmc,avgmc,
essay1,essay2,essay3,essay4,totessay,avgessay,
totscore,avgscore,ma081,ma082,ec011,ec012,ma083,en093$
```

Using these recode and create commands, yields the following relevant variable definitions:

```
recode; size; 0/9=1; 10/19=2; 20/29=3; 30/39=4; 40/49=5;
50/100=6$
create; smallest=size=1; smaller=size=2; small=size=3;
large=size=4; larger=size=5; largest=size=6$
```

TOTALMC: Student's score on 12th grade economics multiple-choice exam (M_i).

AVGMC: Mean multiple-choice score for students at school (\bar{M}_i).

TOTESSAY: Student's score on 12th grade economics essay exam (W_i).

AVGESSAY: Mean essay score for students at school (\bar{W}_i).

ADULTST = 1, if a returning adult student, and 0 otherwise.

SEX = GENDER = 1 if student is female and 0 is male.

ESLFLAG = 1 if English is not student's first language and 0 if it is.

EC011 = 1 if student enrolled in first semester 11 grade economics course, 0 if not.

EN093 = 1 if student was enrolled in ESL English course, 0 if not

MA081 = 1 if student enrolled in the first semester 11 grade math course, 0 if not.

MA082 = 1 if student was enrolled in the second semester 11 grade math course, 0 if not.

MA083 = 1 if student was enrolled in the first semester 12 grade math course, 0 if not.

SMALLER = 1 if student from a school with 10 to 19 test takers, 0 if not.

SMALL = 1 if student from a school with 20 to 29 test takers, 0 if not.

LARGE = 1 if student from a school with 30 to 39 test takers, 0 if not.

LARGER = 1 if student from a school with 40 to 49 test takers, 0 if not.

In all of the regressions, the effect of being at a school with more than 49 test takers is captured in the constant term, against which the other dummy variables are compared. The smallest schools need to be rejected to treat the mean scores as exogenous and unaffected by any individual student's test performance, which is accomplished with the following command:

```
Reject; smallest = 1$
```

The descriptive statistics on the relevant variables are then obtained with the following command, yielding the LIMDEP output table shown:

```
Dstat;RHS=TOTALMC,AVGMC,TOTESSAY,AVGESSAY,ADULTST,SEX,ESLFLAG,
EC01,EN093,MA081,MA082,MA083,SMALLER,SMALL,LARGE,LARGER$
```

```
Descriptive Statistics
All results based on nonmissing observations.
=====
Variable          Mean          Std.Dev.      Minimum      Maximum      Cases
=====
-----
All observations in current sample
-----
TOTALMC    12.4355795    3.96194160    .000000000    20.0000000    3710
AVGMC      12.4355800    1.97263767    6.41666700    17.0714300    3710
TOTESSAY   18.1380054    9.21191366    .000000000    40.0000000    3710
AVGESSAY   18.1380059    4.66807071    5.70000000    29.7857100    3710
ADULTST    .512129380E-02 .713893539E-01 .000000000    1.00000000    3710
SEX        .390566038    .487943012    .000000000    1.00000000    3710
ESLFLAG    .641509434E-01 .245054660    .000000000    1.00000000    3710
EC011      .677088949    .467652064    .000000000    1.00000000    3710
EN093      .622641509E-01 .241667268    .000000000    1.00000000    3710
MA081      .591374663    .491646035    .000000000    1.00000000    3710
MA082      .548787062    .497681208    .000000000    1.00000000    3710
MA083      .420215633    .493659946    .000000000    1.00000000    3710
SMALLER    .462264151    .498641179    .000000000    1.00000000    3710
SMALL      .207277628    .405410797    .000000000    1.00000000    3710
LARGE      .106469003    .308478530    .000000000    1.00000000    3710
LARGER     .978436658E-01 .297143201    .000000000    1.00000000    3710
```

For comparison with the two-stage least squares results, we start with the least squares regressions shown after this paragraph. The least squares estimations are typical of those found in multiple-choice and essay score correlation studies, with correlation coefficients of 0.77 and 0.78. The essay mark or score, W , is the most significant variable in the multiple-choice score regression (first of the two tables) and the multiple-choice mark, M , is the most significant variable in the essay regression (second of the two tables). Results like these have led researchers to conclude that the essay and multiple-choice marks are good predictors of each other. Notice also that both the mean multiple-choice and mean essay marks are significant in their respective equations, suggesting that something in the classroom environment or group experience influences individual test scores. Finally, being female has a significant negative effect on the multiple choice-test score, but a significant positive effect on the essay score, as expected from the least squares results reported by others. We will see how these results hold up in the two-stage least squares regressions.

```
Regress;LHS=TOTALMC;RHS=TOTESSAY,ONE,ADULTST,SEX,AVGMC,
ESLFLAG,EC011,EN093,MA081,MA082,MA083,SMALLER,SMALL,LARGE,LARGER$
```

```
+-----+
| Ordinary least squares regression Weighting variable = none
| Dep. var. = TOTALMC Mean= 12.43557951 , S.D.= 3.961941603
| Model size: Observations = 3710, Parameters = 15, Deg.Fr.= 3695
| Residuals: Sum of squares= 23835.89955 , Std.Dev.= 2.53985
| Fit: R-squared= .590590, Adjusted R-squared = .58904
| Model test: F[ 14, 3695] = 380.73, Prob value = .00000
| Diagnostic: Log-L = -8714.8606, Restricted(b=0) Log-L = -10371.4459
| LogAmemiyaPrCrt.= 1.868, Akaike Info. Crt.= 4.706
| Autocorrel: Durbin-Watson Statistic = 1.99019, Rho = .00490
+-----+
```

```

+-----+
+-----+-----+-----+-----+-----+-----+
|Variable | Coefficient | Standard Error | b/St.Er. | P[|Z|>z] | Mean of X |
+-----+-----+-----+-----+-----+-----+
TOTEISSAY .2707916883 .54725877E-02 49.481 .0000 18.138005
Constant 2.654956801 .33936151 7.823 .0000
ADULTST .4674947703 .59221296 .789 .4299 .51212938E-02
SEX -.5259548390 .91287080E-01 -5.762 .0000 .39056604
AVGMC .3793818833 .25290373E-01 15.001 .0000 12.435580
ESLFLAG .3933259495 .85245570 .461 .6445 .64150943E-01
EC011 .1722643321E-01 .92648817E-01 .186 .8525 .67708895
EN093 -.3117337847 .86493864 -.360 .7185 .62264151E-01
MA081 -.1208070545 .18084020 -.668 .5041 .59137466
MA082 .3827058262 .19467371 1.966 .0493 .54878706
MA083 .3703758129 .11847674 3.126 .0018 .42021563
SMALLER .6721051012E-01 .14743497 .456 .6485 .46226415
SMALL -.5687831831E-02 .15706323 -.036 .9711 .20727763
LARGE .6635816769E-01 .17852633 .372 .7101 .10646900
LARGER .5654860817E-01 .18217561 .310 .7563 .97843666E-01
(Note: E+nn or E-nn means multiply by 10 to + or -nn power.)

```

Regress;LHS=TOTEISSAY;RHS=TOTALMC,ONE, ADULTST,SEX,AVGESSAY,
ESLFLAG,EC011,EN093,MA081,MA082,MA083,SMALLER,SMALL,LARGE,LARGER\$

```

+-----+
| Ordinary least squares regression Weighting variable = none
| Dep. var. = TOTEISSAY Mean= 18.13800539 , S.D.= 9.211913659
| Model size: Observations = 3710, Parameters = 15, Deg.Fr.= 3695
| Residuals: Sum of squares= 123011.3151 , Std.Dev.= 5.76986
| Fit: R-squared= .609169, Adjusted R-squared = .60769
| Model test: F[ 14, 3695] = 411.37, Prob value = .00000
| Diagnostic: Log-L = -11759.0705, Restricted(b=0) Log-L = -13501.8081
| LogAmemiyaPrCrt.= 3.509, Akaike Info. Crt.= 6.347
| Autocorrel: Durbin-Watson Statistic = 2.03115, Rho = -.01557
+-----+

```

```

+-----+-----+-----+-----+-----+-----+
|Variable | Coefficient | Standard Error | b/St.Er. | P[|Z|>z] | Mean of X |
+-----+-----+-----+-----+-----+-----+
TOTALMC 1.408895961 .28223608E-01 49.919 .0000 12.435580
Constant -8.948704180 .55427657 -16.145 .0000
ADULTST -.8291495512 1.3454556 -.616 .5377 .51212938E-02
SEX 1.239956900 .20801072 5.961 .0000 .39056604
AVGESSAY .4000235352 .23711680E-01 16.870 .0000 18.138006
ESLFLAG .4511403830 1.9369352 .233 .8158 .64150943E-01
EC011 .2985371912 .21044864 1.419 .1560 .67708895
EN093 -2.020881931 1.9647001 -1.029 .3037 .62264151E-01
MA081 .8495120566 .41061265 2.069 .0386 .59137466
MA082 .1590915478 .44249860 .360 .7192 .54878706
MA083 1.809541566 .26793945 6.754 .0000 .42021563
SMALLER .6170663022 .33054246 1.867 .0619 .46226415
SMALL .2693408755 .35476913 .759 .4477 .20727763
LARGE .2646447973 .40526280 .653 .5137 .10646900
LARGER .6150288712E-01 .41436703 .148 .8820 .97843666E-01
(Note: E+nn or E-nn means multiply by 10 to + or -nn power.)

```

Theoretical considerations discussed in Module Two, Part One, suggest that these least squares estimates involve a simultaneous equation bias that is brought about by an apparent reverse causality between the two forms of testing. Consistent estimation of the parameters in this simultaneous equation system is possible with two-stage least squares, where our instrument (\hat{M}_i) for M_i is obtained by a least squares regression of M_i on SEX, ADULTST, AVGMC, AVGESSAY, ESLFLAG, SMALLER, SMALL, LARGE, LARGER, EC011, EN093, MA081, MA082, and MA083. Our instrument (\hat{W}_i) for W_i is obtained by a least squares regression of

W_i on SEX, ADULTST, AVGMC, AVGESSAY, ESLFLAG, SMALLER, SMALL, LARGE, LARGER, EC011, EN093, MA081, MA082, and MA083. LIMDEP will do these regressions and the subsequent regressions for M and W employing these instruments via the following commands, which yield the subsequent output:¹

```
2SLS; LHS = TOTALMC; RHS = TOTESSAY, ONE, ADULTST, SEX, AVGMC,
ESLFLAG, EC011, EN093, MA081, MA082, MA083, SMALLER, SMALL, LARGE,
LARGER; INST = ONE, SEX, ADULTST, AVGMC, AVGESSAY, ESLFLAG,
SMALLER, SMALL, LARGE, LARGER, EC011, EN093, MA081, MA082, MA083$
```

```
+-----+
Two stage least squares regression Weighting variable = none
Dep. var. = TOTALMC Mean= 12.43557951 , S.D.= 3.961941603
Model size: Observations = 3710, Parameters = 15, Deg.Fr.= 3695
Residuals: Sum of squares= 46157.78754 , Std.Dev.= 3.53440
Fit: R-squared= .203966, Adjusted R-squared = .20095
      (Note: Not using OLS. R-squared is not bounded in [0,1]
Model test: F[ 14, 3695] = 67.63, Prob value = .00000
Diagnostic: Log-L = -9940.7797, Restricted(b=0) Log-L = -10371.4459
      LogAmemiyaPrCrt.= 2.529, Akaike Info. Crt.= 5.367
Autocorrel: Durbin-Watson Statistic = 2.07829, Rho = -.03914
+-----+
+-----+-----+-----+-----+-----+-----+
|Variable | Coefficient | Standard Error | b/St.Er. | P[|Z|>z] | Mean of X |
+-----+-----+-----+-----+-----+-----+
TOTESSAY - .5247790489E-01 .36407219E-01 -1.441 .1495 18.138005
Constant -.3038295700 .57375703 -.530 .5964
ADULTST .2533493567 .82444633 .307 .7586 .51212938E-02
SEX -.8971949978E-01 .13581404 -.661 .5089 .39056604
AVGMC .9748840572 .74429145E-01 13.098 .0000 12.435580
ESLFLAG .6744471036 1.1866603 .568 .5698 .64150943E-01
EC011 .2925430155 .13244518 2.209 .0272 .67708895
EN093 -1.588715660 1.2118154 -1.311 .1899 .62264151E-01
MA081 .2995655100 .25587578 1.171 .2417 .59137466
MA082 .8159710785 .27507326 2.966 .0030 .54878706
MA083 1.635255739 .21583992 7.576 .0000 .42021563
SMALLER .2715919941 .20639788 1.316 .1882 .46226415
SMALL .4372991271E-01 .21863306 .200 .8415 .20727763
LARGE .1981182700 .24885626 .796 .4260 .10646900
LARGER -.8677104536E-01 .25400196 -.342 .7326 .97843666E-01
2SLS; LHS = TOTESSAY; RHS = TOTALMC, ONE, ADULTST, SEX, AVGE
SSAY, ESLFLAG, EC011, EN093, MA081, MA082, MA083, SMALLER, SMALL,
LARGE, LARGER; INST = ONE, SEX, ADULTST, AVGMC, AVGESSAY, ESLFLAG,
SMALLER, SMALL, LARGE, LARGER, EC011, EN093, MA081, MA082, MA083$
```

```
+-----+
Two stage least squares regression Weighting variable = none
Dep. var. = TOTESSAY Mean= 18.13800539 , S.D.= 9.211913659
Model size: Observations = 3710, Parameters = 15, Deg.Fr.= 3695
Residuals: Sum of squares= 201898.9900 , Std.Dev.= 7.39196
Fit: R-squared= .355924, Adjusted R-squared = .35348
      (Note: Not using OLS. R-squared is not bounded in [0,1]
Model test: F[ 14, 3695] = 145.85, Prob value = .00000
Diagnostic: Log-L = -12678.2066, Restricted(b=0) Log-L = -13501.8081
      LogAmemiyaPrCrt.= 4.005, Akaike Info. Crt.= 6.843
Autocorrel: Durbin-Watson Statistic = 2.10160, Rho = -.05080
+-----+
+-----+-----+-----+-----+-----+-----+
|Variable | Coefficient | Standard Error | b/St.Er. | P[|Z|>z] | Mean of X |
+-----+-----+-----+-----+-----+-----+
TOTALMC .2788777265E-01 .15799711 .177 .8599 12.435580
Constant -1.179740796 1.1193206 -1.054 .2919
ADULTST -.1690793751 1.7252757 -.098 .9219 .51212938E-02
```

SEX	.6854633662	.27355130	2.506	.0122	.39056604
AVGESSAY	.8417622152	.57819872E-01	14.558	.0000	18.138006
ESLFLAG	1.723698602	2.4855173	.693	.4880	.64150943E-01
EC011	.7128702679	.27353325	2.606	.0092	.67708895
EN093	-3.983249481	2.5265144	-1.577	.1149	.62264151E-01
MA081	1.069628788	.52662071	2.031	.0422	.59137466
MA082	1.217026971	.57901457	2.102	.0356	.54878706
MA083	3.892551120	.41430603	9.395	.0000	.42021563
SMALLER	.3348223746	.42463421	.788	.4304	.46226415
SMALL	-.1364832691	.45674848	-.299	.7651	.20727763
LARGE	.3418924354	.51926721	.658	.5103	.10646900
LARGER	-.8251220287E-01	.53110191	-.155	.8765	.97843666E-01

The 2SLS results differ from the least squares results in many ways. The essay mark or score, W , is no longer a significant variable in the multiple-choice regression and the multiple-choice mark, M , is likewise insignificant in the essay regression. Each score appears to be measuring something different when the regressor and error-term-induced bias is eliminated by our instrumental variable estimators.

Both the mean multiple-choice and mean essay scores continue to be significant in their respective equations. But now being female is insignificant in explaining the multiple-choice test score. Being female continues to have a significant positive effect on the essay score.

DURBIN, HAUSMAN AND WU TEST FOR ENDOGENEITY

The theoretical argument is strong for treating multiple-choice and essay scores as endogenous when employed as regressors in the explanation of the other. Nevertheless, this endogeneity can be tested with the Durbin, Hausman and Wu specification test, which is a two-step procedure in LIMDEP versions prior to 9.0.4.ⁱⁱ

Either a Wald statistic, in a Chi-square (χ^2) test with K^* degrees of freedom, or an F statistic with K^* and $n - (K + K^*)$ degrees of freedom, is used to test the joint significance of the contribution of the predicted values ($\hat{\mathbf{X}}^*$) of a regression of the K^* endogenous regressors, in matrix \mathbf{X}^* , on the exogenous variables (and column of ones for the constant term) in matrix \mathbf{Z} :

$$\mathbf{y} = \mathbf{X}\boldsymbol{\beta} + \hat{\mathbf{X}}^* \boldsymbol{\gamma} + \boldsymbol{\varepsilon}^*,$$

where $\mathbf{X}^* = \mathbf{Z}\boldsymbol{\lambda} + \mathbf{u}$, $\hat{\mathbf{X}}^* = \hat{\mathbf{Z}}\hat{\boldsymbol{\lambda}}$, and $\hat{\boldsymbol{\lambda}}$ is a least squares estimator of $\boldsymbol{\lambda}$.

H_o : $\boldsymbol{\gamma} = 0$, the variables in \mathbf{Z} are exogenous

H_A : $\boldsymbol{\gamma} \neq 0$, at least one of the variables in \mathbf{Z} is endogenous

In our case, $K^* = 1$ when the essay score is to be tested as an endogenous regressor in the multiple-choice equation and when the multiple-choice regressor is to be tested as endogenous in the essay equation. $\hat{\mathbf{X}}^*$ is an $n \times 1$ vector of predicted essay scores from a regression of essay scores on all the exogenous variables (for subsequent use in the multiple-choice equation) or an $n \times 1$ vector of predicted multiple-choice scores from a regression of multiple-choice scores on all the exogenous variables (for subsequent use in the essay equation). Because $K^* = 1$, the relevant

test statistic is either the t , with $n - (K + K^*)$ degrees of freedom for small n or the standard normal, for large n .

In LIMDEP, the predicted essay score is obtained by the following command, where the specification “;keep=Essayhat” tells LIMDEP to predict the essay scores and keep them as a variable called “Essayhat”:

```
Regres; lhs= TOTESSAY; RHS= ONE,ADULTST,SEX, AVGESSAY,AVGMC,
ESLFLAG,EC011,EN093,MA081,MA082,MA083, SMALLER, SMALL, LARGE, LARGER
;keep=Essayhat$
```

The predicted essay scores are then added as a regressor in the original multiple-choice regression:

```
Regress; LHS=TOTALMC; RHS=TOTESSAY, ONE, ADULTST, SEX, AVGMC,
ESLFLAG, EC011, EN093, MA081, MA082, MA083, SMALLER, SMALL, LARGE,
LARGER, Essayhat$
```

The test statistic for the Essayhat coefficient is then used in the test of endogeneity. In the below LIMDEP output, we see that the calculated standard normal test statistic z value is -12.916 , which far exceeds the absolute value of the 0.05 percent Type I error critical 1.96 standard normal value. Thus, the null hypothesis of an exogenous essay score as an explanatory variable for the multiple-choice score is rejected. As theorized, the essay score is endogenous in an explanation of the multiple-choice score.

```
--> Regres; lhs= TOTESSAY; RHS= ONE,ADULTST,SEX, AVGESSAY,AVGMC,
ESLFLAG,EC011,EN093,MA081,MA082,MA083, SMALLER, SMALL, LARGE, LARGER;keep=Esayhat$
```

```

+-----+
| Ordinary least squares regression      Weighting variable = none
| Dep. var. = TOTESSAY Mean= 18.13800539 , S.D.= 9.211913659
| Model size: Observations = 3710, Parameters = 15, Deg.Fr.= 3695
| Residuals: Sum of squares= 205968.5911 , Std.Dev.= 7.466609
| Fit: R-squared= .345598, Adjusted R-squared = .34312
| Model test: F[ 14, 3695] = 139.38, Prob value = .00000
| Diagnostic: Log-L = -12715.2253, Restricted(b=0) Log-L = -13501.8081
|              LogAmemiyaPrCrt.= 4.025, Akaike Info. Crt.= 6.863
| Autocorrel: Durbin-Watson Statistic = 2.10143, Rho = -.05072
+-----+

```

Variable	Coefficient	Standard Error	b/St. Er.	P[Z >z]	Mean of X
Constant	-1.186477526	1.1613927	-1.022	.3070	
ADULTST	-.1617772661	1.7412483	-.093	.9260	.51212938E-02
SEX	.6819632415	.27234747	2.504	.0123	.39056604
AVGESSAY	.8405321032	.64642750E-01	13.003	.0000	18.138006
AVGMC	.2714761464E-01	.15534612	.175	.8613	12.435580
ESLFLAG	1.739961011	2.5067133	.694	.4876	.64150943E-01
EC011	.7199749635	.27219191	2.645	.0082	.67708895
EN093	-4.021669541	2.5417647	-1.582	.1136	.62264151E-01
MA081	1.076407689	.53146100	2.025	.0428	.59137466
MA082	1.237970826	.57190601	2.165	.0304	.54878706
MA083	3.932399725	.34253928	11.480	.0000	.42021563
SMALLER	.3418961082	.43385196	.788	.4307	.46226415
SMALL	-.1350660711	.46222353	-.292	.7701	.20727763


```

LARGE      .3469098130      .52477155      .661      .5086      .10646900
LARGER     -.8480793833E-01     .53618108      -.158      .8743      .97843666E-01
(Note: E+nn or E-nn means multiply by 10 to + or -nn power.)

```

```

--> Regress;LHS=TOTALMC;RHS=TOTESAY,ONE,ADULTST,SEX,AVGMC,
    ESLFLAG,EC011,EN093,MA081,MA082,MA083,SMALLER,SMALL,LARGE,LARGER,
    Essayhat$

```

```

+-----+
| Ordinary least squares regression      Weighting variable = none
| Dep. var. = TOTALMC Mean= 12.43557951 , S.D.= 3.961941603
| Model size: Observations = 3710, Parameters = 16, Deg.Fr.= 3694
| Residuals: Sum of squares= 22805.95017 , Std.Dev.= 2.48471
| Fit: R-squared= .608280, Adjusted R-squared = .60669
| Model test: F[ 15, 3694] = 382.41, Prob value = .00000
| Diagnostic: Log-L = -8632.9227, Restricted(b=0) Log-L = -10371.4459
|              LogAmemiyaPrCrt.= 1.825, Akaike Info. Crt.= 4.662
| Autocorrel: Durbin-Watson Statistic = 2.07293, Rho = -.03647
+-----+

```

Variable	Coefficient	Standard Error	b/St.Er.	P[Z >z]	Mean of X
TOTESAY	.2855834321	.54748868E-02	52.162	.0000	18.138005
Constant	-.3038295700	.40335588	-.753	.4513	
ADULTST	.2533493567	.57959250	.437	.6620	.51212938E-02
SEX	-.8971949978E-01	.95478380E-01	-.940	.3474	.39056604
AVGMC	.9748840572	.52324297E-01	18.632	.0000	12.435580
ESLFLAG	.6744471036	.83423185	.808	.4188	.64150943E-01
EC011	.2925430155	.93110045E-01	3.142	.0017	.67708895
EN093	-1.588715660	.85191616	-1.865	.0622	.62264151E-01
MA081	.2995655100	.17988277	1.665	.0958	.59137466
MA082	.8159710785	.19337874	4.220	.0000	.54878706
MA083	1.635255739	.15173722	10.777	.0000	.42021563
SMALLER	.2715919941	.14509939	1.872	.0612	.46226415
SMALL	.4372991270E-01	.15370083	.285	.7760	.20727763
LARGE	.1981182700	.17494798	1.132	.2574	.10646900
LARGER	-.8677104536E-01	.17856546	-.486	.6270	.97843666E-01
ESSAYHAT	-.3380613370	.26173585E-01	-12.916	.0000	18.138005

(Note: E+nn or E-nn means multiply by 10 to + or -nn power.)

The similar estimation routine to test for the endogeneity of the multiple-choice test score in the essay equation yields a calculated z test statistic of -11.713 , which far exceeds the absolute value of its 1.96 critical value. Thus, the null hypothesis of an exogenous multiple-choice score as an explanatory variable for the essay score is rejected. As theorized, the multiple-choice score is endogenous in an explanation of the essay score.

```

--> Regress;LHS=TOTALMC;RHS=ONE,ADULTST,SEX,AVGMC,AVGESSAY,
    ESLFLAG,EC011,EN093,MA081,MA082,MA083,SMALLER,SMALL,LARGE,LARGER;
    keep=MChat$

```

```

+-----+
| Ordinary least squares regression      Weighting variable = none
| Dep. var. = TOTALMC Mean= 12.43557951 , S.D.= 3.961941603
| Model size: Observations = 3710, Parameters = 15, Deg.Fr.= 3695
| Residuals: Sum of squares= 39604.31525 , Std.Dev.= 3.27389
| Fit: R-squared= .319748, Adjusted R-squared = .31717
| Model test: F[ 14, 3695] = 124.06, Prob value = .00000
| Diagnostic: Log-L = -9656.7280, Restricted(b=0) Log-L = -10371.4459
|              LogAmemiyaPrCrt.= 2.376, Akaike Info. Crt.= 5.214
| Autocorrel: Durbin-Watson Statistic = 2.07600, Rho = -.03800
+-----+

```

Variable	Coefficient	Standard Error	b/St.Er.	P[Z >z]	Mean of X
ONE	1.0000000000	0.0000000000	1.000	0.000	18.138005
ADULTST	.2533493567	.57959250	.437	.6620	.51212938E-02
SEX	-.8971949978E-01	.95478380E-01	-.940	.3474	.39056604
AVGMC	.9748840572	.52324297E-01	18.632	.0000	12.435580
AVGESSAY	.6744471036	.83423185	.808	.4188	.64150943E-01
EC011	.2925430155	.93110045E-01	3.142	.0017	.67708895
EN093	-1.588715660	.85191616	-1.865	.0622	.62264151E-01
MA081	.2995655100	.17988277	1.665	.0958	.59137466
MA082	.8159710785	.19337874	4.220	.0000	.54878706
MA083	1.635255739	.15173722	10.777	.0000	.42021563
SMALLER	.2715919941	.14509939	1.872	.0612	.46226415
SMALL	.4372991270E-01	.15370083	.285	.7760	.20727763
LARGE	.1981182700	.17494798	1.132	.2574	.10646900
LARGER	-.8677104536E-01	.17856546	-.486	.6270	.97843666E-01

```

Constant      -.2415657153      .50927203      -.474      .6353
ADULTST      .2618390887      .76353941      .343      .7317      .51212938E-02
SEX          -.1255075019      .11942469     -1.051     .2933      .39056604
AVGMC        .9734594072      .68119457E-01  14.290    .0000      12.435580
AVGESSAY     -.4410936377E-01 .28345921E-01 -1.556    .1197      18.138006
ESLFLAG      .5831375952      1.0991967     .531      .5958      .64150943E-01
EC011        .2547602379      .11935647     2.134     .0328      .67708895
EN093        -1.377666868     1.1145668     -1.236    .2164      .62264151E-01
MA081        .2430778897      .23304627     1.043     .2969      .59137466
MA082        .7510049632      .25078145     2.995     .0027      .54878706
MA083        1.428891640      .15020388     9.513     .0000      .42021563
SMALLER      .2536500026      .19024459     1.333     .1824      .46226415
SMALL        .5081789714E-01 .20268556     .251      .8020      .20727763
LARGE        .1799131698      .23011294     .782      .4343      .10646900
LARGER       -.8232050244E-01 .23511603     -.350     .7262      .97843666E-01

```

```

--> Regress,LHS=TOTESSAY;RHS=TOTALMC,ONE, ADULTST,SEX,AVGESSAY,
    ESLFLAG,EC011,EN093,MA081,MA082,MA083, SMALLER, SMALL,LARGE,LARGER,
    MChat$

```

```

+-----+
| Ordinary least squares regression      Weighting variable = none
| Dep. var. = TOTESSAY Mean= 18.13800539 , S.D.= 9.211913659
| Model size: Observations = 3710, Parameters = 16, Deg.Fr.= 3694
| Residuals: Sum of squares= 118606.0003 , Std.Dev.= 5.66637
| Fit: R-squared= .623166, Adjusted R-squared = .62164
| Model test: F[ 15, 3694] = 407.25, Prob value = .00000
| Diagnostic: Log-L = -11691.4200, Restricted(b=0) Log-L = -13501.8081
|              LogAmemiyaPrCrt.= 3.473, Akaike Info. Crt.= 6.311
| Autocorrel: Durbin-Watson Statistic = 2.09836, Rho = -.04918
+-----+

```

```

+-----+-----+-----+-----+-----+-----+
|Variable | Coefficient | Standard Error |b/St.Er. |P[|Z|>z] | Mean of X|
+-----+-----+-----+-----+-----+-----+
TOTALMC   1.485222426  .28473026E-01  52.162   .0000    12.435580
Constant  -1.179740796  .85802415     -1.375   .1691
ADULTST   -.1690793751  1.3225239     -.128    .8983    .51212938E-02
SEX        .6854633662   .20969294     3.269    .0011    .39056604
AVGESSAY  .8417622152   .44322287E-01  18.992   .0000    18.138006
ESLFLAG   1.723698602   1.9052933     .905     .3656    .64150943E-01
EC011     .7128702679   .20967911     3.400    .0007    .67708895
EN093     -3.983249481  1.9367199     -2.057   .0397    .62264151E-01
MA081     1.069628788   .40368533     2.650    .0081    .59137466
MA082     1.217026971   .44384827     2.742    .0061    .54878706
MA083     3.892551120   .31758961     12.257   .0000    .42021563
SMALLER   .3348223746   .32550676     1.029    .3037    .46226415
SMALL     -.1364832691   .35012421     -.390    .6967    .20727763
LARGE     .3418924354   .39804844     .859     .3904    .10646900
LARGER    -.8251220288E-01 .40712043     -.203    .8394    .97843666E-01
MCHAT     -1.457334653   .12441585     -11.713  .0000    12.435580
(Note: E+nn or E-nn means multiply by 10 to + or -nn power.)

```

CONCLUDING COMMENTS

This cookbook-type introduction to the use of instrumental variables and two-stage least squares regression and testing for endogeneity has just scratched the surface of this controversial problem in statistical estimation and inference. It was intended to enable researchers to begin using instrumental variables in their work and to enable readers of that work to have an idea of what is being done. To learn more about these methods there is no substitute for a graduate level textbook treatment such as that found in William Greene's *Econometric Analysis*.

REFERENCES

Becker, William E. and Carol Johnston (1999). "The Relationship Between Multiple Choice and Essay Response Questions in Assessing Economics Understanding," *Economic Record* (Economic Society of Australia), Vol. 75 (December): 348-357.

Greene, William (2003). *Econometric Analysis*. 5th Edition, New Jersey: Prentice Hall.

ENDNOTES

ⁱ In the default mode, relatively large samples are required for 2SLS in LIMDEP because a routine aimed at providing consistent estimators is employed; thus, for example, no degrees of freedom adjustment is made for variances; *i.e.*,

$$\hat{\sigma}^2 = (1/n) \sum (i^{\text{th}} \text{ prediction error})^2$$

As William Greene states, "this is consistent with most published sources, but (curiously enough) inconsistent with most other commercially available computer programs." The degrees of freedom correction for small samples is obtainable by adding the following specification to the 2SLS command: ;DFC

ⁱⁱIn Limdep version 9.0.4, the following command will automatically test x_3 for endogeneity:

```
Regress; lhs=y; rhs=one,x2,x3; inst=one,x2,x4; Wu test$
```

Because x_3 is not an instrument, LIMDEP knows the test for endogeneity is on this variable.