

PLSC 503 – Spring 2020

Simultaneity and Endogeneity

February 27, 2020

Consider:

$$Y_1 = \mathbf{X}_1\boldsymbol{\beta}_1 + \gamma_1 Y_2 + \mathbf{u}_1$$

$$Y_2 = \mathbf{X}_2\boldsymbol{\beta}_2 + \gamma_2 Y_1 + \mathbf{u}_2$$

Rewrite:

$$\begin{aligned} Y_1 &= \mathbf{X}_1\boldsymbol{\beta}_1 + \gamma_1[\mathbf{X}_2\boldsymbol{\beta}_2 + \gamma_2 Y_1 + \mathbf{u}_2] + \mathbf{u}_1 \\ &= \mathbf{X}_1\boldsymbol{\beta}_1 + \gamma_1(\mathbf{X}_2\boldsymbol{\beta}_2) + \gamma_1\gamma_2 Y_1 + \gamma_1\mathbf{u}_2 + \mathbf{u}_1 \\ Y_1 - \gamma_1\gamma_2 Y_1 &= \mathbf{X}_1\boldsymbol{\beta}_1 + \gamma_1(\mathbf{X}_2\boldsymbol{\beta}_2) + \gamma_1\mathbf{u}_2 + \mathbf{u}_1 \\ (1 - \gamma_1\gamma_2)Y_1 &= \mathbf{X}_1\boldsymbol{\beta}_1 + \gamma_1(\mathbf{X}_2\boldsymbol{\beta}_2) + \gamma_1\mathbf{u}_2 + \mathbf{u}_1 \\ Y_1 &= \mathbf{X}_1 \left(\frac{1}{1 - \gamma_1\gamma_2} \boldsymbol{\beta}_1 \right) + \mathbf{X}_2 \left(\frac{\gamma_1}{1 - \gamma_1\gamma_2} \boldsymbol{\beta}_2 \right) + \left(\frac{\gamma_1\mathbf{u}_2 + \mathbf{u}_1}{1 - \gamma_1\gamma_2} \right) \\ &= \Delta_1\mathbf{X}_1 + \Delta_2\mathbf{X}_2 + \mathbf{e} \end{aligned}$$

$$Y_1 = \mathbf{x}_1 \left(\frac{1}{1 - \gamma_1 \gamma_2} \beta_1 \right) + \mathbf{x}_2 \left(\frac{\gamma_1}{1 - \gamma_1 \gamma_2} \beta_2 \right) + \left(\frac{\gamma_1 \mathbf{u}_2 + \mathbf{u}_1}{1 - \gamma_1 \gamma_2} \right)$$

means

$$\frac{\partial Y_1}{\partial X_\ell} = \frac{\beta_\ell}{1 - \gamma_1 \gamma_2}.$$

But

$$\hat{\Delta}_1 \neq \hat{\beta}_1.$$

For (e.g.)

$$Y_1 = \mathbf{X}_1\beta_1 + \gamma_1 Y_2 + \mathbf{u}_1$$

we have:

$$E(Y_2, \mathbf{u}_1) = \frac{\gamma_2}{1 - \gamma_1\gamma_2} \sigma_{\mathbf{u}}^2$$

Result:

- Bias (unless $\gamma_2 = 0$)
- Inconsistency

- OLS
- Lagged Variables
- Two-Stage Least Squares (2SLS)
- Systems of Equations / 3SLS / etc.

Recall that a simple linear model:

$$\mathbf{Y} = \mathbf{X}\beta + \mathbf{u}$$

gives us:

$$\hat{\beta}_{OLS} = \beta + (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{u}.$$

Suppose $\text{Cov}(\mathbf{X}, \mathbf{u}) \neq \mathbf{0}$, but we have \mathbf{Z} with

- $\text{Cov}(\mathbf{Z}, \mathbf{X}) \neq \mathbf{0}$ and
- $\text{Cov}(\mathbf{Z}, \mathbf{u}) = \mathbf{0}$.

Then:

$$\begin{aligned}\hat{\beta}_{IV} &= (\mathbf{Z}'\mathbf{X})^{-1}\mathbf{Z}'\mathbf{Y} \\ &= (\mathbf{Z}'\mathbf{X})^{-1}\mathbf{Z}'(\mathbf{X}\beta + \mathbf{u}) \\ &= \beta + (\mathbf{Z}'\mathbf{X})^{-1}\mathbf{Z}'\mathbf{u}\end{aligned}$$

is consistent.

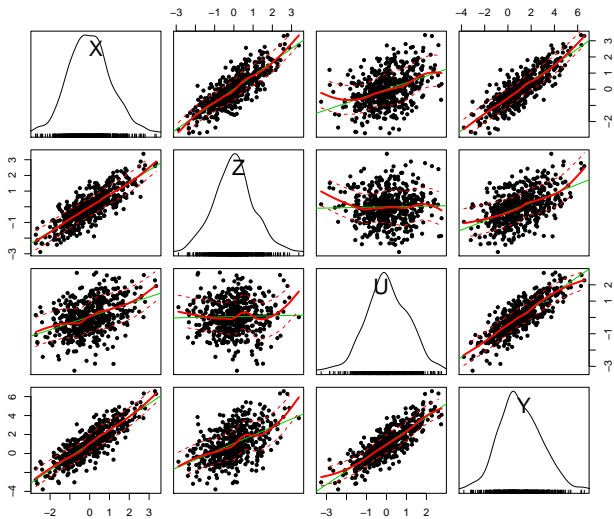
- Regress endogenous \mathbf{X} s variables on $\{\mathbf{Z}, \mathbf{X}\}$
- Generate $\hat{\mathbf{X}}$ s
- Regress Y on $\hat{\mathbf{X}}$ to get β_{2SLS} .
- Adjust standard error estimates

```
library(MASS)
library(sem)
library(car)

seed<-1337
set.seed(seed)

mu<-c(0,0,0) # <== X, Z, U
Sigma<-matrix(c(1,0.8,0.4,0.8,1,0,0.4,0,1),
              nrow=3,byrow=TRUE)          # Cor(X,Y)=0.8, etc.
Vars<- mvrnorm(500,mu,Sigma)
colnames(Vars)<-c("X","Z","U")
Vars<-data.frame(Vars)

Vars$Y<- 1 + Vars$X + Vars$U
```

Plain Old OLS...

```
> OLS<- lm(Y~X,data=Vars)
> summary(OLS)
```

Call:

```
lm(formula = Y ~ X, data = Vars)
```

Residuals:

Min	1Q	Median	3Q	Max
-3.3809	-0.6058	-0.0102	0.6320	2.9470

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.04770	0.04209	24.89	<2e-16 ***
X	1.40254	0.04005	35.02	<2e-16 ***

Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.9413 on 498 degrees of freedom

Multiple R-squared: 0.7112, Adjusted R-squared: 0.7106

F-statistic: 1226 on 1 and 498 DF, p-value: < 2.2e-16

Two-Stage Least Squares

```
> TSLS<-tsls(Y~I(X),data=Vars,instruments=~Z)
> summary(TSLS)
```

2SLS Estimates

Model Formula: $Y \sim I(X)$

Instruments: $\sim Z$

Residuals:

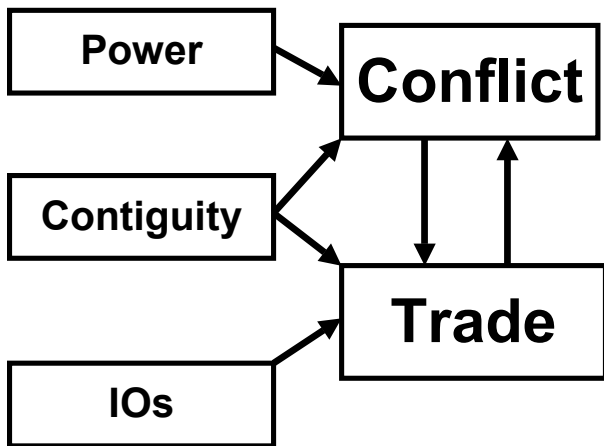
Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-3.29300	-0.68210	-0.06139	0.00000	0.76270	2.70300

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.0491828	0.0456017	23.00754	< 2.22e-16 ***
I(X)	1.0302012	0.0536909	19.18763	< 2.22e-16 ***

Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 1.0196738 on 498 degrees of freedom

IV: A (Toy) Example



```
> summary(IRData)
```

dyadid	logdisputes	logtrade	I0s
Min. : 2020	Min. : -0.6931	Min. : -0.6931	Min. : 4.579
1st Qu.: 135155	1st Qu.: -0.6931	1st Qu.: 2.4079	1st Qu.: 19.500
Median : 220484	Median : -0.6931	Median : 5.5786	Median : 27.704
Mean : 275526	Mean : -0.2627	Mean : 4.6518	Mean : 30.891
3rd Qu.: 385710	3rd Qu.: 0.0000	3rd Qu.: 7.1248	3rd Qu.: 39.289
Max. : 900920	Max. : 3.4965	Max. : 11.5037	Max. : 93.700

contiguity	capratio	GDPgrowth
Min. : 0.0000	Min. : 1.081	Min. : -9.0800
1st Qu.: 0.0000	1st Qu.: 4.849	1st Qu.: -0.2923
Median : 0.0000	Median : 26.577	Median : 0.8363
Mean : 0.3207	Mean : 196.310	Mean : 0.5097
3rd Qu.: 1.0000	3rd Qu.: 144.035	3rd Qu.: 1.7106
Max. : 1.0000	Max. : 7451.982	Max. : 7.0460

```
> OLSwar<-lm(logdisputes~logtrade+contiguity+capratio,data=IRData)
> summary(OLSwar)
```

Call:

```
lm(formula = logdisputes ~ logtrade + contiguity + capratio,
    data = IRData)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.828	-0.326	-0.269	-0.090	3.455

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.4253192	0.0602014	-7.06	3.5e-12 ***
logtrade	0.0085581	0.0105739	0.81	0.419
contiguity	0.4622674	0.0712406	6.49	1.5e-10 ***
capratio	-0.0001296	0.0000647	-2.00	0.045 *

Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.853 on 813 degrees of freedom

Multiple R-squared: 0.083, Adjusted R-squared: 0.0796

F-statistic: 24.5 on 3 and 813 DF, p-value: 3.35e-15

2SLS “By-Hand” (stage one)

```
> ITrade<-lm(logtrade~contiguity+IOs+capratio)
> summary(ITrade)
```

Residuals:

Min	1Q	Median	3Q	Max
-6.0385	-1.7666	0.4139	1.6154	7.6029

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	0.7319793	0.1912570	3.827	0.000140	***
contiguity	1.3386037	0.1816041	7.371	4.17e-13	***
IOs	0.1218373	0.0055313	22.027	< 2e-16	***
capratio	-0.0013913	0.0001626	-8.555	< 2e-16	***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.239 on 813 degrees of freedom

Multiple R-squared: 0.5535, Adjusted R-squared: 0.5519

F-statistic: 335.9 on 3 and 813 DF, p-value: < 2.2e-16

2SLS “By-Hand” (stage two)

```
> IVWarByHand<-with(IRData, lm(logdisputes~capratio+contiguity+  
+                               (ITrade$fitted.values)))  
> summary(IVWarByHand)
```

Call:

```
lm(formula = logdisputes ~ capratio + contiguity + (ITrade$fitted.values))
```

Residuals:

Min	1Q	Median	3Q	Max
-1.006	-0.362	-0.278	-0.049	3.530

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.1515180	0.0832287	-1.82	0.06905 .
capratio	-0.0002664	0.0000705	-3.78	0.00017 ***
contiguity	0.6263774	0.0788444	7.94	6.5e-15 ***
ITrade\$fitted.values	-0.0558374	0.0171921	-3.25	0.00121 **

Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1

Residual standard error: 0.848 on 813 degrees of freedom

Multiple R-squared: 0.094, Adjusted R-squared: 0.0907

F-statistic: 28.1 on 3 and 813 DF, p-value: <2e-16

2SLS, Automagically

```
> library(sem)
> TwoSLSWar<-tsls(logdisputes~contiguity+capratio+I(logtrade),
  instruments=~contiguity+capratio+IOs)
> summary(TwoSLSWar)
```

2SLS Estimates

Model Formula: logdisputes ~ contiguity + capratio + I(logtrade)

Instruments: ~contiguity + capratio + IOs

Residuals:

	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
	-1.21e+00	-5.24e-01	-2.26e-01	-7.44e-17	-2.10e-02	3.65e+00

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.1515180	8.562e-02	-1.770	7.717e-02
contiguity	0.6263774	8.111e-02	7.722	3.353e-14
capratio	-0.0002664	7.252e-05	-3.674	2.543e-04
I(logtrade)	-0.0558374	1.769e-02	-3.157	1.652e-03

Residual standard error: 0.8723 on 813 degrees of freedom

Weak Instruments

```
> OLSTrade<-lm(logtrade~logdisputes+contiguity+IOs)
> summary(OLSTrade)
```

Residuals:

Min	1Q	Median	3Q	Max
-6.2467	-2.2067	0.4275	1.6659	6.1264

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	0.191111	0.182875	1.045	0.296
logdisputes	0.408116	0.095067	4.293	1.98e-05 ***
contiguity	1.357557	0.193109	7.030	4.38e-12 ***
IOs	0.133778	0.005614	23.831	< 2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.312 on 813 degrees of freedom

Multiple R-squared: 0.5241, Adjusted R-squared: 0.5223

F-statistic: 298.4 on 3 and 813 DF, p-value: < 2.2e-16

Weak Instruments (continued)

```
> TwoSLSTrade<-tsls(logtrade~contiguity+IOs+I(logdisputes),  
  instruments=~contiguity+capratio+IOs)  
> summary(TwoSLSTrade)
```

2SLS Estimates

Model Formula: logtrade ~ contiguity + IOs + I(logdisputes)

Instruments: ~contiguity + capratio + IOs

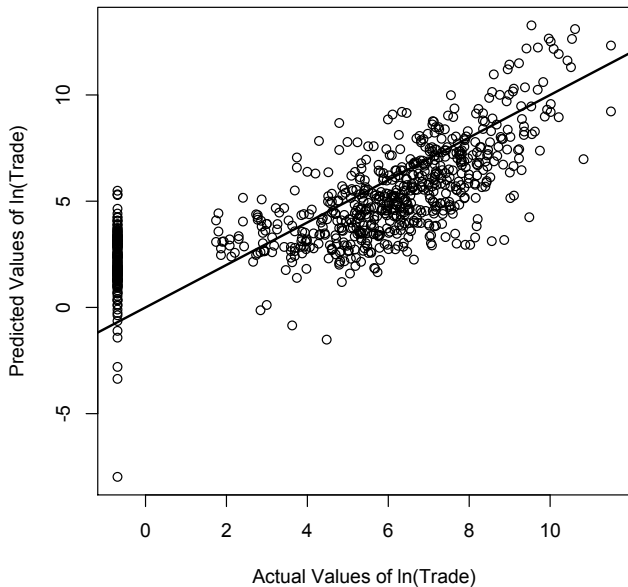
Residuals:

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
-2.57e+01	-1.46e+00	1.36e+00	2.84e-14	4.00e+00	1.09e+01

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.150	0.85122	2.526	1.173e-02
contiguity	-2.728	1.52615	-1.787	7.427e-02
IOs	0.172	0.02045	8.408	2.220e-16
I(logdisputes)	7.371	2.45198	3.006	2.727e-03

Residual standard error: 6.3721 on 813 degrees of freedom

Pretty Good Instrument (Trade)



Crappy Instrument (War)

