# PLSC 503 – Spring 2017 Simultaneity and Endogeneity

March 22, 2017

# Endogeneity

Consider:

$$\begin{aligned} \mathbf{Y}_1 &= \mathbf{X}_1 \boldsymbol{\beta}_1 + \gamma_1 \mathbf{Y}_2 + \mathbf{u}_1 \\ \\ \mathbf{Y}_2 &= \mathbf{X}_2 \boldsymbol{\beta}_2 + \gamma_2 \mathbf{Y}_1 + \mathbf{u}_2 \end{aligned}$$

Rewrite:

$$\begin{array}{rcl} Y_1 & = & \textbf{X}_1\beta_1 + \gamma_1[\textbf{X}_2\beta_2 + \gamma_2\,Y_1 + \textbf{u}_2] + \textbf{u}_1 \\ & = & \textbf{X}_1\beta_1 + \gamma_1(\textbf{X}_2\beta_2) + \gamma_1\gamma_2\,Y_1 + \gamma_1\textbf{u}_2 + \textbf{u}_1 \\ Y_1 - \gamma_1\gamma_2\,Y_1 & = & \textbf{X}_1\beta_1 + \gamma_1(\textbf{X}_2\beta_2) + \gamma_1\textbf{u}_2 + \textbf{u}_1 \\ (1 - \gamma_1\gamma_2)Y_1 & = & \textbf{X}_1\beta_1 + \gamma_1(\textbf{X}_2\beta_2) + \gamma_1\textbf{u}_2 + \textbf{u}_1 \\ Y_1 & = & \textbf{X}_1\left(\frac{1}{1 - \gamma_1\gamma_2}\beta_1\right) + \textbf{X}_2\left(\frac{\gamma_1}{1 - \gamma_1\gamma_2}\beta_2\right) + \left(\frac{\gamma_1\textbf{u}_2 + \textbf{u}_1}{1 - \gamma_1\gamma_2}\right) \\ & = & \Delta_1\textbf{X}_1 + \Delta_2\textbf{X}_2 + \textbf{e} \end{array}$$

### "Reduced Form"

$$\mathbf{\textit{Y}}_{1} = \mathbf{\textit{X}}_{1}\left(\frac{1}{1-\gamma_{1}\gamma_{2}}\boldsymbol{\beta}_{1}\right) + \mathbf{\textit{X}}_{2}\left(\frac{\gamma_{1}}{1-\gamma_{1}\gamma_{2}}\boldsymbol{\beta}_{2}\right) + \left(\frac{\gamma_{1}\mathbf{\textit{u}}_{2} + \mathbf{\textit{u}}_{1}}{1-\gamma_{1}\gamma_{2}}\right)$$

means

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

But

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

# Simultaneity Bias

For (e.g.)

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

we have:

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

### What To Do

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

$$Y = X\beta + u$$

has

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

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

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

Then

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

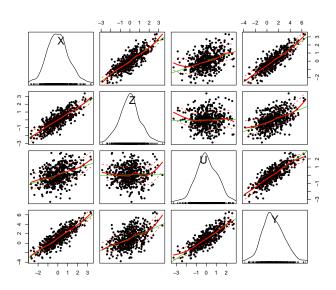
is consistent.

# 2SLS: How-To

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

### **IV** Estimation

```
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) #
Vars<- mvrnorm(500,mu,Sigma)</pre>
colnames(Vars)<-c("X","Z","U")</pre>
Vars<-data.frame(Vars)</pre>
Vars$Y<- 1 + Vars$X + Vars$U
```



- > 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

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

2SLS Estimates

Model Formula: Y ~ I(X)

Instruments: ~Z

#### Residuals:

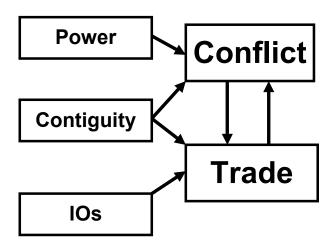
Min. 1st Qu. Median Mean 3rd Qu. Max. -3.29300 -0.68210 -0.06139 0.00000 0.76270 2.70300

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
	conrotio	CDParouth	
contiguity	capiacio	GDI GI OW CII	
Min. :0.0000	Min. : 1.081	0	
0 0	-	Min. :-9.0800	
Min. :0.0000	Min. : 1.081	Min. :-9.0800	
Min. :0.0000 1st Qu.:0.0000	Min.: 1.081 1st Qu.: 4.849	Min. :-9.0800 1st Qu.:-0.2923	
Min. :0.0000 1st Qu.:0.0000 Median :0.0000	Min. : 1.081 1st Qu.: 4.849 Median : 26.577 Mean : 196.310	Min. :-9.0800 1st Qu.:-0.2923 Median : 0.8363 Mean : 0.5097	

- > OLSWar<-lm(logdisputes~logtrade+contiguity+capratio)
- > summary(OLSWar)

#### Residuals:

Min 1Q Median 3Q Max -0.82840 -0.32644 -0.26860 -0.08972 3.45504

#### Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) -4.253e-01 6.020e-02 -7.065 3.46e-12 \*\*\*
logtrade 8.558e-03 1.057e-02 0.809 0.4185
contiguity 4.623e-01 7.124e-02 6.489 1.50e-10 \*\*\*
capratio -1.296e-04 6.467e-05 -2.003 0.0455 \*
--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: 0.08301, Adjusted R-squared: 0.07962 F-statistic: 24.53 on 3 and 813 DF, p-value: 3.345e-15

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

## "By-Hand"

- > 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

### "By-Hand"

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

#### Residuals:

```
Min 1Q Median 3Q Max
-1.0055 -0.3618 -0.2782 -0.0492 3.5301
```

#### Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) -1.515e-01 8.323e-02 -1.821 0.069050 .
capratio -2.664e-04 7.049e-05 -3.780 0.000168 ***
contiguity 6.264e-01 7.884e-02 7.944 6.49e-15 ***
ITrade$fitted.values -5.584e-02 1.719e-02 -3.248 0.001210 **
---
Signif. codes: 0 *** 0.001 ** 0.01 * 0.05 . 0.1 1
```

Residual standard error: 0.8479 on 813 degrees of freedom Multiple R-squared: 0.09402, Adjusted R-squared: 0.09068 F-statistic: 28.12 on 3 and 813 DF, p-value: < 2.2e-16

### 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:

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

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

#### 2SLS Estimates

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

Instruments: ~contiguity + capratio + IOs

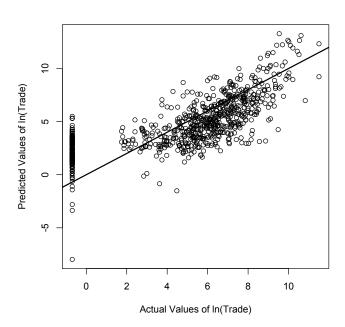
#### Residuals:

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)

