

Assignment Simultaneous Equations Model

Simultaneous Equations System

$$y_{1t} = \beta_{10} + \beta_{11}y_{2t} + \beta_{12}x_{1t} + \beta_{13}x_{2t} + u_{1t} \quad (1)$$

$$y_{2t} = \beta_{20} + \beta_{21}y_{1t} + \beta_{22}x_{3t} + \beta_{23}x_{4t} + u_{2t} \quad (2)$$

Endogenous variables in this system include y_{1t} and y_{2t}

Exogenous variables in this system include x_{1t} , x_{2t} , x_{3t} , and x_{4t}

1. State reduce form model of these system models.
2. Estimate reduce form model using OLS and prediction of the endogenous variables.
3. Estimate structural form using predicted endogenous variables as independent variables in the structural form model.
4. Estimate the structural models of these system equations using OLS, 2SLS, and 3SLS. Concerning on the asymptotic property, which estimated results are the most appropriated results? Why?

1. State reduce form model of these system models.

$$y_{1t} = \pi_{10} + \pi_{11}x_{1t} + \pi_{12}x_{2t} + \pi_{13}x_{3t} + \pi_{14}x_{4t} + w_{1t}$$

$$y_{2t} = \pi_{20} + \pi_{21}x_{1t} + \pi_{22}x_{2t} + \pi_{23}x_{3t} + \pi_{24}x_{4t} + w_{2t}$$

2. Estimate reduce form model using OLS and prediction of the endogenous variables.

```
. reg y1 x1 x2 x3 x4
```

Source	SS	df	MS	Number of obs	=	500
Model	7700218.45	4	1925054.61	F(4, 495)	=	189.18
Residual	5037024.74	495	10175.8076	Prob > F	=	0.0000
				R-squared	=	0.6045
				Adj R-squared	=	0.6013
Total	12737243.2	499	25525.5375	Root MSE	=	100.88

y1	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	.7351842	.0458451	16.04	0.000	.6451092	.8252591
x2	-1.370992	.1062249	-12.91	0.000	-1.579699	-1.162284
x3	-.3716883	.1571014	-2.37	0.018	-.6803562	-.0630205
x4	.9451346	.050141	18.85	0.000	.8466191	1.04365
_cons	-.9002973	5.618986	-0.16	0.873	-11.9403	10.13971

```
. predict y1hat, xb
```

```
. reg y2 x*
```

Source	SS	df	MS	Number of obs	=	500
Model	36037999.6	4	9009499.91	F(4, 495)	=	716.26
Residual	6226340.32	495	12578.4653	Prob > F	=	0.0000
				R-squared	=	0.8527
				Adj R-squared	=	0.8515
Total	42264340	499	84698.0761	Root MSE	=	112.15

y2	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x1	.265324	.0509709	5.21	0.000	.165178	.36547
x2	-.5918622	.1181016	-5.01	0.000	-.8239044	-.35982
x3	-1.292387	.1746665	-7.40	0.000	-1.635566	-.9492077
x4	2.946954	.0557471	52.86	0.000	2.837424	3.056484
_cons	-2.189755	6.247227	-0.35	0.726	-14.46411	10.0846

```
. predict y2hat, xb
```

3. Estimate structural form using predicted endogenous variables as independent variables in the structural form model.

```
. reg y1 y2hat x1 x2
```

Source	SS	df	MS	Number of obs	=	500
Model	7699473.59	3	2566491.2	F(3, 496)	=	252.69
Residual	5037769.6	496	10156.7936	Prob > F	=	0.0000
				R-squared	=	0.6045
				Adj R-squared	=	0.6021
Total	12737243.2	499	25525.5375	Root MSE	=	100.78

y1	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y2hat	.3201675	.0168776	18.97	0.000	.287007	.3533279
x1	.6506053	.0459881	14.15	0.000	.5602498	.7409609
x2	-1.1816	.1060642	-11.14	0.000	-1.389991	-.9732098
_cons	-.0825759	5.588965	-0.01	0.988	-11.06354	10.89839

```
. reg y2 y1hat x3 x4
```

Source	SS	df	MS	Number of obs	=	500
Model	36032833.3	3	12010944.4	F(3, 496)	=	956.02
Residual	6231506.64	496	12563.5215	Prob > F	=	0.0000
				R-squared	=	0.8526
				Adj R-squared	=	0.8517
Total	42264340	499	84698.0761	Root MSE	=	112.09

y2	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
y1hat	.3887183	.0540226	7.20	0.000	.2825769 .4948596
x3	-1.149046	.1751029	-6.56	0.000	-1.493081 -.8050109
x4	2.578087	.0743161	34.69	0.000	2.432074 2.724101
_cons	-.4182315	5.814539	-0.07	0.943	-11.8424 11.00593

4. Estimate the structural models of these system equations using OLS, 2SLS, and 3SLS. Concerning on the asymptotic property, which estimated results are the most appropriated results? Why?

```
. reg y1 y2 x1 x2
```

Source	SS	df	MS	Number of obs	=	500
Model	10726508.6	3	3575502.87	F(3, 496)	=	881.99
Residual	2010734.57	496	4053.90034	Prob > F	=	0.0000
				R-squared	=	0.8421
				Adj R-squared	=	0.8412
Total	12737243.2	499	25525.5375	Root MSE	=	63.67

y1	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
y2	.399428	.0098383	40.60	0.000	.3800981 .4187579
x1	.6301685	.0290345	21.70	0.000	.5731227 .6872144
x2	-1.161171	.0669997	-17.33	0.000	-1.29281 -1.029533
_cons	-5.839926	3.518283	-1.66	0.098	-12.7525 1.072649

```
. est store m1
```

```
. reg y2 y1 x3 x4
```

Source	SS	df	MS	Number of obs	=	500
Model	40607309.2	3	13535769.7	F(3, 496)	=	4051.67
Residual	1657030.81	496	3340.78792	Prob > F	=	0.0000
				R-squared	=	0.9608
				Adj R-squared	=	0.9606
Total	42264340	499	84698.0761	Root MSE	=	57.8

y2	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
y1	.7478655	.0189107	39.55	0.000	.7107106 .7850204
x3	-1.050503	.0901201	-11.66	0.000	-1.227568 -.8734393
x4	2.250549	.033475	67.23	0.000	2.184778 2.316319
_cons	-17.91048	2.827999	-6.33	0.000	-23.46682 -12.35415

```
. est store m2
```

```
. ivregress 2sls y1 x1 x2 (y2=x1 x2 x3 x4)
```

Instrumental variables (2SLS) regression	Number of obs	=	500
	wald chi2(3)	=	1693.05
	Prob > chi2	=	0.0000
	R-squared	=	0.8215
	Root MSE	=	67.437

y1	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
y2	.3201675	.0112935	28.35	0.000	.2980326 .3423023
x1	.6506053	.0307725	21.14	0.000	.5902922 .7109184

```

      x2 |      -1.1816   .0709719   -16.65   0.000   -1.320703   -1.042498
    _cons |      -.082576   3.739805    -0.02   0.982    -7.41246    7.247308
-----+-----

```

```

Instrumented:  y2
Instruments:   x1 x2 x3 x4

```

```
. est store m3
```

```
. ivregress 2sls y2 x3 x4 (y1=x1 x2 x3 x4)
```

```

Instrumental variables (2SLS) regression          Number of obs   =       500
                                                wald chi2(3)     =     6295.02
                                                Prob > chi2       =       0.0000
                                                R-squared         =     0.9323
                                                Root MSE         =     75.657

```

```

-----+-----
      y2 |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      y1 |   .3887183   .0364645    10.66   0.000     .3172491     .4601874
      x3 |  -1.149046   .1181921    -9.72   0.000    -1.380698    -.9173935
      x4 |   2.578087   .0501624    51.39   0.000     2.479771     2.676404
    _cons |  -.4182316   3.924735    -0.11   0.915    -8.110571     7.274108
-----+-----

```

```

Instrumented:  y1
Instruments:   x3 x4 x1 x2

```

```
. est store m4
```

```
. reg3 (y1 y2 x1 x2) (y2 y1 x3 x4), inst(x1 x2 x3 x4) 3sls
```

```
Three-stage least-squares regression
```

```

-----+-----
Equation      Obs     Parms      RMSE    "R-sq"     chi2       P
-----+-----
y1             500        3     67.50756   0.8211    1694.59   0.0000
y2             500        3     75.63899   0.9323    6295.31   0.0000
-----+-----

```

```

-----+-----
      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
y1
      y2 |   .3198315   .0112879    28.33   0.000     .2977076     .3419554
      x1 |   .6650667   .0266978    24.91   0.000     .6127399     .7173935
      x2 |  -1.140358   .055968    -20.38   0.000    -1.250054    -1.030663
    _cons |   .9128191   3.58841     0.25   0.799    -6.120335     7.945973
-----+-----
y2
      y1 |   .3890072   .0364571    10.67   0.000     .3175526     .4604618
      x3 |  -1.185279   .0738911    -16.04   0.000    -1.330103    -1.040455
      x4 |   2.576427   .0499839    51.55   0.000     2.47846     2.674394
    _cons |  -.3246522   3.917497    -0.08   0.934    -8.002806     7.353501
-----+-----

```

```

Endogenous variables:  y1 y2
Exogenous variables:   x1 x2 x3 x4

```

```
. est store m5
```

```
. est table m1 m2 m3 m4 m5, se stat(N rss rss_1 rss_2 rmse rmse_1 rmse_2 F chi2
chi2_1 chi2_2 r2 r2_1 r2_2 r2_a)
```

```

-----+-----
Variable      m1          m2          m3          m4          m5
-----+-----
-
      y2 |   .39942802          .32016746
          |   .00983831          .0112935
      x1 |   .63016855          .6506053
          |   .02903452          .03077254
      x2 |  -1.1611715        -1.1816004
          |   .06699971          .07097193
      y1 |          .74786547          .38871825
          |          .01891067          .03646451
      x3 |          -1.0505035        -1.1490458
          |          .09012008          .1181921
-----+-----

```

	x4		2.2505488		2.5780874
	_cons	-5.8399263	.03347504	-.08257597	.05016238
		3.5182829	-17.910485	3.7398053	-.41823164
			2.8279987		3.924735
y1					
	y2				.31983146
					.01128791
	x1				.6650667
					.02669784
	x2				-1.1403583
					.05596803
	_cons				.91281906
					3.58841
y2					
	y1				.38900721
					.03645709
	x3				-1.1852792
					.07389113
	x4				2.5764269
					.0499839
	_cons				-.32465224
					3.9174972
Statistics					
	N	500	500	500	500
	rss	2010734.6	1657030.8	2273850.3	2862011.9
	rss_1				2278635.4
	rss_2				2860628.6
	rmse	63.670247	57.799549	67.436641	75.657279
	rmse_1				67.50756
	rmse_2				75.638993
	F	881.99082	4051.6699		
	chi2			1693.0476	6295.0181
	chi2_1				1694.5862
	chi2_2				6295.3051
	r2	.84213738	.96079364	.82148019	.93228306
	r2_1				.82110451
	r2_2				.93231579
	r2_a	.84118257	.96055651	.82040043	.93187348

legend: b/se

	y1			y2		
	OLS	2SLS	3SLS	OLS	2SLS	3SLS
y1				0.7479 (0.0189)	0.3887 (0.0365)	0.3890 (0.0365)
y2	0.3994 (0.0098)	0.3202 (0.0113)	0.3198 (0.0113)			
x1	0.6302 (0.0290)	0.6506 (0.0308)	0.6651 (0.0267)			
x2	-1.1612 (0.0670)	-1.1816 (0.0710)	-1.1404 (0.0560)			
x3				-1.0505 (0.0901)	-1.1490 (0.1182)	-1.1853 (0.0739)
x4				2.2505 (0.0335)	2.5781 (0.0502)	2.5764 (0.0500)
Constant	-5.8399 (3.5183)	-0.0826 (3.7398)	0.9128 (3.5884)	-17.9105 (2.8280)	-0.4182 (3.9247)	-0.3247 (3.9175)
Obs.	500	500	500	500	500	500
RSS	2010735	2273850	2278635	1657031	2862012	2860629
RMSE	63.67	67.44	67.51	57.80	75.66	75.64
F-test	881.99			4051.67		
Chi2- test		1693.05	1694.59		6295.02	6295.31
R- squares	0.8421	0.8215	0.8211	0.9608	0.9323	0.9323
Adj. R2	0.8412	0.8204		0.9606	0.9319	

Note: Value in parenthesis is Standard Error (S.E.)

Data Generating Process

```

set obs 500
g t=_n
tsset t
g x1=rnormal(10,100)
g x2=rnormal(-30,45)
g x3=rnormal(1,30)
g x4=rnormal(15,90)
matrix C = (1, .8 \ .8, 1)
corr2data u1 u2, n(500) means(0 0) sds(70 75) corr(C)
g y1=1 in 1
g y2=2 in 1
forvalue t=1(1)500 {
  replace y1=0.5+0.7*x1-1.2*x2+u1 in `t'
  replace y2=0.5-1.1*x3+2.5*x4+u2 in `t'
  replace y1=0.5+0.3*y2+0.7*x1-1.2*x2+u1 in `t'
  replace y2=0.5+0.4*y1-1.1*x3+2.5*x4+u2 in `t'
  replace y1=0.5+0.3*y2+0.7*x1-1.2*x2+u1 in `t'
  replace y2=0.5+0.4*y1-1.1*x3+2.5*x4+u2 in `t'
  replace y1=0.5+0.3*y2+0.7*x1-1.2*x2+u1 in `t'
  replace y2=0.5+0.4*y1-1.1*x3+2.5*x4+u2 in `t'
}

```

* Compare 2SLS vs GMM

```
. reg y1 y2hat x1 x2
```

Source	SS	df	MS	Number of obs	=	500
Model	7699473.59	3	2566491.2	F(3, 496)	=	252.69
Residual	5037769.6	496	10156.7936	Prob > F	=	0.0000
				R-squared	=	0.6045
				Adj R-squared	=	0.6021
Total	12737243.2	499	25525.5375	Root MSE	=	100.78

y1	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
y2hat	.3201675	.0168776	18.97	0.000	.287007	.3533279
x1	.6506053	.0459881	14.15	0.000	.5602498	.7409609
x2	-1.1816	.1060642	-11.14	0.000	-1.389991	-.9732098
_cons	-.0825759	5.588965	-0.01	0.988	-11.06354	10.89839

```
. ivregress 2sls y1 x1 x2 (y2=x1 x2 x3 x4)
```

Instrumental variables (2SLS) regression	Number of obs	=	500
	wald chi2(3)	=	1693.00
	Prob > chi2	=	0.0000
	R-squared	=	0.8215
	Root MSE	=	67.437

y1	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
y2	.3201675	.0112935	28.35	0.000	.2980326	.3423023
x1	.6506053	.0307725	21.14	0.000	.5902922	.7109184
x2	-1.1816	.0709719	-16.65	0.000	-1.320703	-1.042498
_cons	-.082576	3.739805	-0.02	0.982	-7.41246	7.247308

```
Instrumented: y2
Instruments: x1 x2 x3 x4
```

```
. ivregress gmm y1 x1 x2 (y2=x1 x2 x3 x4)
```

Instrumental variables (GMM) regression	Number of obs	=	500
	wald chi2(3)	=	1596.25
	Prob > chi2	=	0.0000
	R-squared	=	0.8216
GMM weight matrix: Robust	Root MSE	=	67.405

y1	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
y2	.3204869	.0116394	27.53	0.000	.2976741	.3432998
x1	.6507274	.0298274	21.82	0.000	.5922668	.709188
x2	-1.18233	.0695007	-17.01	0.000	-1.318549	-1.046111
_cons	-.0439821	3.491438	-0.01	0.990	-6.887074	6.79911

```
Instrumented: y2
Instruments: x1 x2 x3 x4
```

```
.
```