

Assignment - Solution

Simultaneous Equations Model

Export-Import Equations System

$$\ln Ex_t = \beta_{10} + \beta_{11} \ln Im_t + \beta_{12} \ln P_{Ext} + \beta_{13} \ln Fx_t + \beta_{14} \ln GDP_{wt} + \varepsilon_{1t} \quad (1)$$

$$\ln Im_t = \beta_{20} + \beta_{21} \ln Ex_t + \beta_{22} \ln Fx_t + \beta_{23} \ln Inf_t + \varepsilon_{2t} \quad (2)$$

where: $\ln Ex_t$ = Natural log of export at time t
 $\ln Im_t$ = Natural log of Import at time t
 $\ln P_{Ext}$ = Natural log of Export price at time t
 $\ln Fx_t$ = Natural log of Exchange rate at time t
 $\ln GDP_{wt}$ = Natural log of World GDP at time t
 $\ln Inf_t$ = Natural log of Inflation at time t

Endogenous variables in this system include $\ln Ex_t$ and $\ln Im_t$

Exogenous variables in this system include $\ln P_{Ext}$, $\ln Fx_t$, $\ln GDP_{wt}$, and $\ln Inf_t$

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, 3SLS, and I3SLS. Concerning on the asymptotic property, which model is the most appropriated model? Why?

1. State reduce form model of these system models.

$$\ln Ex_t = \pi_{10} + \pi_{11} \ln P_{Ext} + \pi_{12} \ln Fx_t + \pi_{13} \ln GDP_{wt} + \pi_{14} \ln Inf_t + w_{1t}$$

$$\ln Im_t = \pi_{20} + \pi_{21} \ln P_{Ext} + \pi_{22} \ln Fx_t + \pi_{23} \ln GDP_{wt} + \pi_{24} \ln Inf_t + w_{2t}$$

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

```
. reg lnEx lnPx lnFx lnGDP lnInf
```

Source	SS	df	MS	Number of obs = 37		
Model	12.8567881	4	3.21419703	F(4, 32)	= 66.92	
Residual	1.5370373	32	.048032416	Prob > F	= 0.0000	
				R-squared	= 0.8932	
				Adj R-squared	= 0.8799	
				Root MSE	= .21916	
Total	14.3938254	36	.399828484			

lnEx	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnPx	.3233985	.1490153	2.17	0.038	.0198642	.6269328
lnFx	-.4574346	.2424967	-1.89	0.068	-.9513843	.0365151
lnGDP	1.297045	.1301112	9.97	0.000	1.032018	1.562073
lnInf	-.2376303	.1754565	-1.35	0.185	-.5950235	.1197629
_cons	.9587734	1.468087	0.65	0.518	-2.031621	3.949168

```
. predict lnExhat, xb
```

```
. reg lnIm lnPx lnFx lnGDP lnInf
```

Source	SS	df	MS	Number of obs = 37		
Model	24.1162307	4	6.02905768	F(4, 32) = 20.26		
Residual	9.52343322	32	.297607288	Prob > F = 0.0000		
				R-squared = 0.7169		
				Adj R-squared = 0.6815		
Total	33.6396639	36	.934435109	Root MSE = .54553		

lnIm	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnPx	-.0040172	.3709245	-0.01	0.991	-.7595656	.7515313
lnFx	-1.72784	.6036156	-2.86	0.007	-2.957365	-.4983155
lnGDP	-.5231396	.3238689	-1.62	0.116	-1.182839	.1365598
lnInf	-.3201353	.4367411	-0.73	0.469	-1.209748	.5694772
_cons	26.363	3.654317	7.21	0.000	18.9194	33.8066

```
. predict lnImhat, xb
```

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

```
. reg lnEx lnImhat lnPx lnFx lnGDP
```

Source	SS	df	MS	Number of obs = 37		
Model	12.8567874	4	3.21419685	F(4, 32)	= 66.92	
Residual	1.537038	32	.048032438	Prob > F	= 0.0000	
				R-squared	= 0.8932	
				Adj R-squared	= 0.8799	
				Root MSE	= .21916	
Total	14.3938254	36	.399828484			

lnEx	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lnImhat	.7422783	.5480703	1.35	0.185	-.3741043	1.858661
lnPx	.3263801	.1498666	2.18	0.037	.0211117	.6316484
lnFx	.825103	1.069916	0.77	0.446	-1.354245	3.004451
lnGDP	1.68536	.349491	4.82	0.000	.9734704	2.39725
_cons	-18.60991	14.48383	-1.28	0.208	-48.11251	10.89268

```
. reg lnIm lnExhat lnFx lnInf
```

Source	SS	df	MS	Number of obs =	37
Model	24.0659823	3	8.02199409	F(3, 33) =	27.65
Residual	9.57368167	33	.290111566	Prob > F =	0.0000
				R-squared =	0.7154
				Adj R-squared =	0.6895
Total	33.6396639	36	.934435109	Root MSE =	.53862

lnIm	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnExhat	-.4785147	.1677628	-2.85	0.007	-.8198307 -.1371987
lnFx	-1.845126	.5135224	-3.59	0.001	-2.889896 -.8003571
lnInf	-.3660743	.3992639	-0.92	0.366	-1.178383 .4462342
_cons	27.95926	2.467356	11.33	0.000	22.93939 32.97913

4. Estimate the structural models of these system equations using OLS, 2SLS, 3SLS, and I3SLS. Concerning on the asymptotic property, which model is the most appropriated model? Why?

```
. reg3 (lnEx lnIm lnPx lnFx lnGDP) (lnIm lnEx lnFx lnInf), ols inst(lnPx lnFx lnGDP lnInf)
```

Multivariate regression

Equation	Obs	Parms	RMSE	"R-sq"	F-Stat	P
lnEx	37	4	.1803998	0.9276	102.57	0.0000
lnIm	37	3	.5859579	0.6632	21.66	0.0000

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnEx					
lnIm	.245524	.0579727	4.24	0.000	.1297445 .3613034
lnPx	.2730116	.1136106	2.40	0.019	.0461156 .4999077
lnFx	-.1237281	.2129469	-0.58	0.563	-.5490123 .3015562
lnGDP	1.389122	.107585	12.91	0.000	1.17426 1.603984
_cons	-5.549768	1.944813	-2.85	0.006	-9.433827 -1.66571
lnIm					
lnEx	-.2257254	.1702545	-1.33	0.190	-.5657471 .1142962
lnFx	-2.100195	.554702	-3.79	0.000	-3.208011 -.9923784
lnInf	-.3362366	.4342847	-0.77	0.442	-1.203563 .5310899
_cons	25.30874	2.594182	9.76	0.000	20.1278 30.48968

```
. reg3 (lnEx lnIm lnPx lnFx lnGDP) (lnIm lnEx lnFx lnInf), 2sls inst(lnPx lnFx lnGDP lnInf)
```

Two-stage least-squares regression

Equation	Obs	Parms	RMSE	"R-sq"	F-Stat	P
lnEx	37	4	.3274401	0.7616	29.98	0.0000
lnIm	37	3	.6052138	0.6407	21.90	0.0000

	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
lnEx					
lnIm	.742281	.818843	0.91	0.368	-.8930613 2.377623
lnPx	.3263804	.2239079	1.46	0.150	-.1207946 .7735554
lnFx	.8251084	1.598506	0.52	0.607	-2.367328 4.017545
lnGDP	1.685362	.5221561	3.23	0.002	.6425441 2.72818
_cons	-18.60998	21.63953	-0.86	0.393	-61.82711 24.60714
lnIm					
lnEx	-.4785146	.1885046	-2.54	0.014	-.8549843 -.1020449
lnFx	-1.845126	.5770131	-3.20	0.002	-2.997501 -.6927517
lnInf	-.3660742	.4486279	-0.82	0.417	-1.262046 .5298976
_cons	27.95926	2.772414	10.08	0.000	22.42236 33.49615

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Endogenous variables: lnEx lnIm
Exogenous variables: lnPx lnFx lnGDP lnInf
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```
. reg3 (lnEx lnIm lnPx lnFx lnGDP) (lnIm lnEx lnFx lnInf), 3sls inst(lnPx lnFx
lnGDP lnInf)
```

Three-stage least-squares regression

Equation	Obs	Parms	RMSE	"R-sq"	chi2	P
lnEx	37	4	.3421059	0.6992	138.86	0.0000
lnIm	37	3	.5715641	0.6407	73.67	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
lnEx						
lnIm	.8276964	.7297008	1.13	0.257	-.6024909	2.257884
lnPx	.3875424	.1379849	2.81	0.005	.1170969	.6579879
lnFx	.9316971	1.461526	0.64	0.524	-1.932842	3.796236
lnGDP	1.769392	.4357714	4.06	0.000	.915296	2.623488
_cons	-21.32074	18.90018	-1.13	0.259	-58.36442	15.72293
lnIm						
lnEx	-.4785146	.1780238	-2.69	0.007	-.8274349	-.1295943
lnFx	-1.845126	.5449313	-3.39	0.001	-2.913172	-.7770806
lnInf	-.3660742	.4236843	-0.86	0.388	-1.19648	.4643318
_cons	27.95926	2.618269	10.68	0.000	22.82755	33.09097

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Endogenous variables: lnEx lnIm
Exogenous variables: lnPx lnFx lnGDP lnInf
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```
. reg3 (lnEx lnIm lnPx lnFx lnGDP) (lnIm lnEx lnFx lnInf), 3sls ireg3 inst(lnPx
lnFx lnGDP lnInf)
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Iteration 1: tolerance = .1382335
Iteration 2: tolerance = .02305974
Iteration 3: tolerance = .00427982
Iteration 4: tolerance = .00080918
Iteration 5: tolerance = .00015352
Iteration 6: tolerance = .00002915
Iteration 7: tolerance = 5.534e-06
Iteration 8: tolerance = 1.051e-06
Iteration 9: tolerance = 1.997e-07
```

Three-stage least-squares regression, iterated

Equation	Obs	Parms	RMSE	"R-sq"	chi2	P
lnEx	37	4	.3511994	0.6829	104.53	0.0000
lnIm	37	3	.5715641	0.6407	73.67	0.0000

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
lnEx						
lnIm	.847716	.8361021	1.01	0.311	-.7910139	2.486446
lnPx	.4018776	.1435903	2.80	0.005	.1204458	.6833094
lnFx	.9566794	1.681353	0.57	0.569	-2.338712	4.252071
lnGDP	1.789087	.4936624	3.62	0.000	.8215268	2.756648
_cons	-21.95609	21.5847	-1.02	0.309	-64.26133	20.34916
lnIm						
lnEx	-.4785146	.1780238	-2.69	0.007	-.8274349	-.1295943
lnFx	-1.845126	.5449313	-3.39	0.001	-2.913172	-.7770806
lnInf	-.3660742	.4236843	-0.86	0.388	-1.19648	.4643318
_cons	27.95926	2.618269	10.68	0.000	22.82755	33.09097

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Endogenous variables: lnEx lnIm
Exogenous variables: lnPx lnFx lnGDP lnInf
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