

```
. *a)
. reg loggdp risk
```

Source	SS	df	MS	Number of obs	=	64
				F(1, 62)	=	68.17
Model	36.2163143	1	36.2163143	Prob > F	=	0.0000
Residual	32.9382286	62	.531261752	R-squared	=	0.5237
				Adj R-squared	=	0.5160
Total	69.1545429	63	1.09769116	Root MSE	=	.72888

loggdp	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
risk	.516187	.0625186	8.26	0.000	.3912141	.6411599
_cons	4.687415	.417441	11.23	0.000	3.852962	5.521867

```
. reg risk logmort0
```

Source	SS	df	MS	Number of obs	=	64
				F(1, 62)	=	23.34
Model	37.1754537	1	37.1754537	Prob > F	=	0.0000
Residual	98.7466676	62	1.59268819	R-squared	=	0.2735
				Adj R-squared	=	0.2618
Total	135.922121	63	2.15749399	Root MSE	=	1.262

risk	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
logmort0	-.6132892	.1269412	-4.83	0.000	-.8670411	-.3595374
_cons	9.365895	.6105941	15.34	0.000	8.145335	10.58646

```
. ivregress 2sls loggdp (risk=logmort0)
```

Instrumental variables 2SLS regression	Number of obs	=	64
	Wald chi2(1)	=	36.60
	Prob > chi2	=	0.0000
	R-squared	=	0.1880
	Root MSE	=	.93672

loggdp	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
risk	.9294897	.1536318	6.05	0.000	.6283768	1.230603
_cons	1.994296	1.007904	1.98	0.048	.0188405	3.969751

```
Instrumented: risk
Instruments: logmort0
```

a)

```
. *d)
. *Two stage approach for 2sls
. reg risk logmort0
```

Source	SS	df	MS	Number of obs	=	64
				F(1, 62)	=	23.34
Model	37.1754537	1	37.1754537	Prob > F	=	0.0000
Residual	98.7466676	62	1.59268819	R-squared	=	0.2735
				Adj R-squared	=	0.2618
Total	135.922121	63	2.15749399	Root MSE	=	1.262

risk	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
logmort0	-.6132892	.1269412	-4.83	0.000	-.8670411	-.3595374
_cons	9.365895	.6105941	15.34	0.000	8.145335	10.58646

```
. predict risk_hat, xb
. reg loggdp risk_hat
```

Source	SS	df	MS	Number of obs	=	64
				F(1, 62)	=	53.77
Model	32.1177716	1	32.1177716	Prob > F	=	0.0000
Residual	37.0367713	62	.597367279	R-squared	=	0.4644
				Adj R-squared	=	0.4558
Total	69.1545429	63	1.09769116	Root MSE	=	.7729

loggdp	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
risk_hat	.9294897	.126763	7.33	0.000	.676094	1.182885
_cons	1.994296	.8316306	2.40	0.020	.3318896	3.656701

Yes, they are the same

f)

```
. reg loggdp risk latitude africa
```

Source	SS	df	MS	Number of obs	=	64
Model	45.8984283	3	15.2994761	F(3, 60)	=	39.47
Residual	23.2561146	60	.387601911	Prob > F	=	0.0000
				R-squared	=	0.6637
				Adj R-squared	=	0.6469
Total	69.1545429	63	1.09769116	Root MSE	=	.62258

loggdp	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
risk	.3765228	.0608421	6.19	0.000	.2548205	.4982252
latitude	1.382463	.6440401	2.15	0.036	.0941905	2.670735
africa	-.7232696	.1712967	-4.22	0.000	-1.065914	-.3806251
_cons	5.652234	.4152858	13.61	0.000	4.821539	6.482929

The coefficients of latitude and africa are significant. They can be predictive (show causality) if they are backed by theory.

g)

```
. ivregress 2sls loggdp latitude africa (risk= logmort0)
```

```
Instrumental variables 2SLS regression
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Number of obs	=	64
Wald chi2(3)	=	57.55
Prob > chi2	=	0.0000
R-squared	=	0.3922
Root MSE	=	.81039

loggdp	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
risk	.799968	.2497417	3.20	0.001	.3104831	1.289453
latitude	-.0553109	1.161701	-0.05	0.962	-2.332203	2.221581
africa	-.3479258	.3062581	-1.14	0.256	-.9481806	.252329
_cons	2.99507	1.581523	1.89	0.058	-.1046587	6.094798

```
Instrumented: risk
```

```
Instruments: latitude africa logmort0
```

Now latitude and africa become insignificant.

i)

```
. gen logmort02= (logmort0)^2
. reg risk logmort0 logmort02
```

Source	SS	df	MS	Number of obs	=	64
Model	51.1839123	2	25.5919562	F(2, 61)	=	18.42
Residual	84.738209	61	1.38915097	Prob > F	=	0.0000
				R-squared	=	0.3766
				Adj R-squared	=	0.3561
Total	135.922121	63	2.15749399	Root MSE	=	1.1786

risk	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
logmort0	-2.645684	.6508988	-4.06	0.000	-3.947237	-1.344132
logmort02	.2101141	.066166	3.18	0.002	.077807	.3424211
_cons	13.94859	1.551696	8.99	0.000	10.84579	17.0514

A unit change in logmort0 leads to a **larger** change in risk as compared to the linear reduced form equation.

```
. *structural equation
. ivregress 2sls loggdp (risk=logmort0 logmort02)
```

Instrumental variables 2SLS regression

Number of obs	=	64
Wald chi2(1)	=	46.68
Prob > chi2	=	0.0000
R-squared	=	0.3948
Root MSE	=	.80865

loggdp	Coefficient	Std. err.	z	P> z	[95% conf. interval]
risk	.7722554	.1130303	6.83	0.000	.55072 .9937908
_cons	3.018849	.7434204	4.06	0.000	1.561772 4.475926

Instrumented: risk
Instruments: logmort0 logmort02

We see that coefficient reduces a little. R squared value falls.

j)

```
. estat endogenous
```

Tests of endogeneity
H0: Variables are exogenous

Durbin (score) chi2(1)	=	10.4601	(p = 0.0012)
Wu-Hausman F(1,61)	=	11.9176	(p = 0.0010)

P value is less than 0.05. Thus we reject the null. This implies that there is endogeneity.

k)

```
test logmort0 logmort02
```

(1) logmort0 = 0
(2) logmort02 = 0

F(2, 61)	=	18.42
Prob > F	=	0.0000

Instruments are relevant as F value is greater than 10

l)

```
. ivregress 2sls loggdp (risk=logmort0 logmort02)
```

```
Instrumental variables 2SLS regression      Number of obs   =          64
                                           Wald chi2(1)    =          46.68
                                           Prob > chi2     =          0.0000
                                           R-squared       =          0.3948
                                           Root MSE       =          .80865
```

loggdp	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
risk	.7722554	.1130303	6.83	0.000	.55072	.9937908
_cons	3.018849	.7434204	4.06	0.000	1.561772	4.475926

Instrumented: risk

Instruments: logmort0 logmort02

```
. predict eps_hat, residual
```

```
. reg eps_hat logmort0 logmort02
```

Source	SS	df	MS	Number of obs	=	64
Model	3.35810086	2	1.67905043	F(2, 61)	=	2.66
Residual	38.4926828	61	.631027586	Prob > F	=	0.0780
				R-squared	=	0.0802
				Adj R-squared	=	0.0501
Total	41.8507836	63	.664298153	Root MSE	=	.79437

eps_hat	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
logmort0	.7516194	.4386953	1.71	0.092	-.125606	1.628845
logmort02	-.0876735	.0445948	-1.97	0.054	-.1768463	.0014993
_cons	-1.464118	1.045818	-1.40	0.167	-3.555361	.6271237