

Problem Set 3.

①. $y_1 = \delta_{12}y_2 + \delta_{13}y_3 + \delta_{11}z_1 + \delta_{13}z_3 + u_1$ (1)

$y_2 = \delta_{21}y_1 + \delta_{21}z_1 + u_2$ (2)

$y_3 = \delta_{31}z_1 + \delta_{32}z_2 + \delta_{33}z_3 + \delta_{34}z_4 + u_3$ (3)

for eqn (1) $B_1 = (\delta_{12} \delta_{13} \delta_{11} \delta_{12} \delta_{13} \delta_{14})'$ Note $G=3, M=4$

by normalization; $B_1 = (-1 \delta_{12} \delta_{13} \delta_{11} \delta_{12} \delta_{13} \delta_{14})'$; Restrictions on (1) are $\delta_{12}=0$ and $\delta_{14}=0$

\therefore we have $J_1 = 2$ and $G-1 = 3-1 = 2$; since $J_1 \geq G-1$ order condition is satisfied; ✓

equation (1) might be identified i.e. we need to check further, the rank condition.

Define restriction matrix $R_1 (J_1 \times (G+3))$ such that $R_1 B_1 = 0$.

$\therefore R_1 = \begin{bmatrix} 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$, $B_1 = (-1 \delta_{12} \delta_{13} \delta_{11} \delta_{12} \delta_{13} \delta_{14})'$

and $B = \begin{bmatrix} -1 & \delta_{21} & \delta_{31} \\ \delta_{12} & -1 & \delta_{32} \\ \delta_{13} & \delta_{23} & -1 \\ \delta_{11} & \delta_{21} & \delta_{31} \\ \delta_{12} & \delta_{22} & \delta_{32} \\ \delta_{13} & \delta_{23} & \delta_{33} \\ \delta_{14} & \delta_{24} & \delta_{34} \end{bmatrix}$; $R_1 B = \begin{bmatrix} \delta_{12} & \delta_{22} & \delta_{32} \\ \delta_{14} & \delta_{24} & \delta_{34} \end{bmatrix}$
 $= \begin{bmatrix} 0 & 0 & \delta_{32} \\ 0 & 0 & \delta_{34} \end{bmatrix}$

$\therefore \text{Rank}(R_1 B) = 1 \neq G-1$; Rank condition fails.

\therefore Equation (1) is unidentified. ✓

for eqn (2) $B_2 = (\delta_{21} \delta_{22} \delta_{23} \delta_{21} \delta_{22} \delta_{23} \delta_{24})'$

by normalization; $B_2 = (\delta_{21} -1 \delta_{23} \delta_{21} \delta_{22} \delta_{23} \delta_{24})'$ Restrictions on (2) are $\delta_{23}=0$,

$\delta_{22}=0, \delta_{23}=0, \delta_{24}=0$ \therefore we have $J_2 = 4$ and $G-1 = 3-1 = 2$; since $J_2 \geq G-1$ order cond.

is satisfied, equation (2) might be identified. Further check on Rank cond. is required.

Define restriction matrix $R_2 (J_2 + (G+M))$ such that $R_2 B_2 = 0$.

$\therefore R_2 = \begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}_{4 \times 7}$ and $B = \begin{bmatrix} -1 & \delta_{21} & \delta_{31} \\ \delta_{12} & -1 & \delta_{32} \\ \delta_{13} & \delta_{23} & -1 \\ \delta_{11} & \delta_{21} & \delta_{31} \\ \delta_{12} & \delta_{22} & \delta_{32} \\ \delta_{13} & \delta_{23} & \delta_{33} \\ \delta_{14} & \delta_{24} & \delta_{34} \end{bmatrix}_{7 \times 3}$

$\therefore R_2 B = \begin{bmatrix} \delta_{13} & \delta_{23} & -1 \\ \delta_{12} & \delta_{22} & \delta_{32} \\ \delta_{13} & \delta_{23} & \delta_{33} \\ \delta_{14} & \delta_{24} & \delta_{34} \end{bmatrix} = \begin{bmatrix} \delta_{13} & 0 & -1 \\ 0 & 0 & \delta_{32} \\ \delta_{13} & 0 & \delta_{33} \\ 0 & 0 & \delta_{34} \end{bmatrix}$ ✓

$\therefore \text{Rank}(R_2 B) = 2 = G-1$; Rank condition is satisfied and since $J_2 = 4 > G-1$, Eqn (2) is over identified. ✓

those exogenous

for eqn (2) variables that can be used for instrumenting y_1 variables that help

endogenous variable y_1 but excluded i.e. z_1 and z_3 . ✓

(21) Consider equation (4)

First, we check order condition. In order to pass order condition, there must be at least one exogenous variable is excluded from the equation (4). Since, $\log(\text{land})$ is excluded from equation (4), therefore, equation (4) pass order condition.

Second, we check rank condition. In order to pass rank condition, at least one of exogenous variables excluded from the equation (4) must have a non-zero population coefficient. We regress open on all exogenous variables,

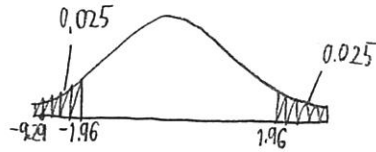
$$\text{open} = 177.0845 + 0.5465 \log(\text{pcinc}) - 7.5677 \log(\text{land}) \quad (a)$$

(15.8483)
(1.4932)
(0.8742)

$$H_0: \beta_{\log(\text{land})} = 0$$

$$H_a: \beta_{\log(\text{land})} \neq 0$$

$$t = \frac{-7.5677}{0.8742} = -9.29$$



Since, the computed t-value falls into rejection region. We reject H_0 at 0.05 significance level. That is, coefficient $\log(\text{land})$ is not equal to zero. So, equation (4) pass rank condition.

In conclusion, equation (4) is identified.

Consider equation (5)

First, we check order condition. Since, there is no exogenous variable which excludes from equation (5), equation isn't identified.

Next, we want to check whether $\log(\text{land})$ can be used as an IV for open

If we assume that $\log(\text{land})$ doesn't correlate with error term in equation (a) and $\text{cov}(\text{open}, \log(\text{land}))$ is not equal to zero, we can conclude that $\log(\text{land})$ can be used as IV for open

We regress equation (4) by using $\log(\text{land})$ as IV for open, we get

$$\text{inf} = 26.8993 - 0.3775 \text{open} + 0.3758 \log(\text{pcinc})$$

(15.7972)
(0.7122)
(7.988)

22.1) check whether eqn (4) is identified.; examn eqn (5)

Order condition: $\log(\text{land})$; there's at least one exovar excluded from eqn(4)

Rank condition: $\beta_{22} \neq 0$ i.e. need to check if there's at least one of the

exo excluded from eqn (4) and have nonzero pop. coefficient in eqn (5)

To check rank condition, we obtain the reduced form for open:

$$\text{open} = \beta_{20} + \alpha_2 (\beta_{10} + \alpha_1 \text{open} + \beta_{11} \log(\text{pcinc}) + u_1) + \beta_2 \log(\text{pcinc}) + \beta_{22} \log(\text{land}) + u_2$$

$$(1 - \alpha_2 \alpha_1) \text{open} = (\beta_{20} + \alpha_1 \beta_{10}) + (\alpha_2 \beta_{11} + \beta_{12}) \log(\text{pcinc}) + \beta_{22} \log(\text{land}) + (\alpha_2 u_1 + u_2)$$

Reduced Form. \Rightarrow

$$\therefore \text{open} = \underbrace{\frac{\beta_{20} + \alpha_1 \beta_{10}}{1 - \alpha_2 \alpha_1}}_{\pi_0} + \underbrace{\frac{\alpha_2 \beta_{11} + \beta_{12}}{1 - \alpha_2 \alpha_1}}_{\pi_1} \log(\text{pcinc}) + \underbrace{\frac{\beta_{22}}{1 - \alpha_2 \alpha_1}}_{\pi_2} \log(\text{land}) + \underbrace{\frac{(\alpha_2 u_1 + u_2)}{(1 - \alpha_2 \alpha_1)}}_{v_1}$$

and test for π_2 i.e. $H_0: \pi_2 = 0$; $H_a: \pi_2 \neq 0$ since test-statistic for π_2 is -9.29 (result attached)

and p-value = .000, we can conclude that π_2 is statistically significant at any

degree of confident level. Thereby rank condition is also satisfied. Note that

with π_2 statistically significant, $\log(\text{land})$ can also be used as an iv for open.

endogen. var, open, with assumption that $\text{cov}(\log(\text{land}), u_1) = 0$.

$$2.2) \text{ OLS: } \hat{\text{inf}} = 25.234 - 0.215 \text{ open}$$

(4.102) (1.093)

$$\text{IV: } \hat{\text{inf}} = 29.607 - .333 \text{ open}$$

(5.61) (1.139)

From stata output reported in table 3, we can see that after dropping $\ln \text{pcinc}$ (which is not statistically significant in IV model in 2.1 ($p\text{-value} = .85$, $t\text{-statistic} = .19$)), coefficient of open is still statistically significant at 0.05 significance level but insignificant at 0.01 level for both IV and OLS. However, when we consider the change in magnitude, OLS and IV estimators give us quite different results. Using IV, we have $\hat{\alpha}_1 = -.333$ indicating that if import as % of GDP increases by 1 percentage point, we will expect an average decrease of ^{annual} inflation of .333 percentage point _{ceteris paribus}. This result is quite similar to $\hat{\alpha}_1 = -.337$ produced by IV before we drop $\ln \text{pcinc}$. This is not surprising as we drop a insignificant ^{exogenous} variable from the model which should not affect our results significantly. However, when we use OLS to estimate the model instead, we have $\hat{\alpha}_1 = -.215$; 1 percentage point increase in import as % of GDP leads to .215 percentage point decrease in annual inflation, on average, *ceteris paribus*. Magnitude of our estimates changes quite a lot signifying that endogeneity problem may be present in OLS model making our OLS estimator inconsistent (we need to conduct endogeneity test to confirm this). Our conclusions regarding statistical significance of our coefficients remain unchanged, however, probably because ~~the~~ standard errors in IV are higher than in OLS.

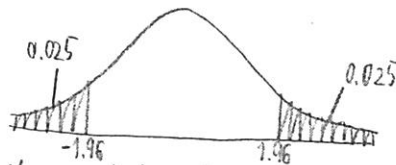
$$(2.3) \text{ inf} = 24.0089 - 0.3370 \text{ open} + 0.8033 \log(\text{pcinc}) - 6.5557 \text{ oil}$$

(15.75165) (0.1419) (2.0804) (9.6279)

$$H_0: \beta_{\text{oil}} = 0$$

$$H_a: \beta_{\text{oil}} \neq 0$$

$$t = \frac{-6.5557}{9.6279} = -0.68$$



Since, the computed t -value doesn't fall into rejection region. We don't reject H_0 at 0.05 significance level. That is, being an oil producer has no effect on inflation, *ceteris paribus*.

alternative:

The computed t -statistic does not fall into rejection region

We cannot reject H_0 at 0.1 level of significance.

There is not enough evidence to conclude that

being an oil producer has a statistically significant partial effect on inflation rate, holding other explanatory variable fixed. It does not have *ceteris paribus* effect on inflation at any reasonable significant level

③ 3.1 Since each equation has a behavioral and the decisions come from 2 agents. As equation (6), the income of each person that the employer will give depends on cigs, educ and age and equation (7), the consumption on cigarettes depends on the decision of ^{each} person, determined by $\log(\text{income})$, educ, age, $\log(\text{cigpric})$, restaurn. Since the decisions come from 2 agents then this is "SEM" and then we can ceteris paribus interpret each equation.

We expect that the sign of γ_5 is negative because the increase in ^{a pack of} cigarette price should reduce the consumption on cigarette per day.

We expect that the sign of γ_6 is negative because the person lives in a state with restaurant smoking restriction should smoke less than who not lives in a state with restaurant restriction.

✓

2) ① Order condition: there are 2 exogenous variable excluded from eq(6) which are "log(cigpric)" and "restaurn" ✓

② Rank condition: "log(cigpric)" and "restaurn" are statistically significant in eq(7)

∴ Assume $\gamma_5 \neq 0$ and or or $\gamma_6 \neq 0$, and ✓

$\text{Cov}(\log(\text{cigpric}), u_1) = 0$, $\text{Cov}(\text{restaurn}, u_1) = 0$

①, ②, and the assumptions stated are satisfied, Eq(6) is identified

$$3.3) \text{ Cigs} = \delta_0 + \delta_1 (\beta_0 + \beta_1 \text{cigs} + \beta_2 \text{educ} + \beta_3 \text{age} + \beta_4 \text{age}^2 + u_1) + \delta_2 \text{educ} + \delta_3 \text{age} + \delta_4 \text{age}^2 + \delta_5 \ln \text{cigprice} + \delta_6 \text{resta}$$

$$\rightarrow \text{L Cigs} = \frac{\delta_0 + \delta_1 \beta_0}{1 - \delta_1 \beta_1} + \frac{\delta_1 \beta_2 + \delta_2}{1 - \delta_1 \beta_1} \text{educ} + \frac{\delta_1 \beta_3 + \delta_3}{1 - \delta_1 \beta_1} \text{age} + \frac{\delta_1 \beta_4 + \delta_4}{1 - \delta_1 \beta_1} \text{age}^2 + \frac{\delta_5}{1 - \delta_1 \beta_1} \ln \text{cigprice} + \frac{\delta_6}{1 - \delta_1 \beta_1} \text{restaurn}$$

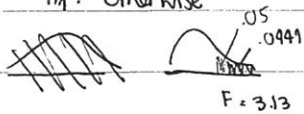
Reduced Form:

$$\text{Cigs} = \pi_0 + \pi_1 \text{educ} + \pi_2 \text{age} + \pi_3 \text{age}^2 + \pi_4 \ln \text{cigprice} + \pi_5 \text{restaurn} + v$$

$$\hat{\text{Cigs}} = 1.58 - .45 \text{educ} + .823 \text{age} - .0095 \text{age}^2 - .351 \ln \text{cigprice} - 2.736 \text{restaurn}$$

$$H_0: \pi_4 = \pi_5 = 0$$

$$H_1: \text{otherwise}$$



According to stata output, F-statistic = 3.13 falls into rejection

region. We reject $H_0: \pi_4 = \pi_5 = 0$ at 0.05 level of significance

and conclude that $\ln \text{cigprice}$ and restaurn can jointly statistically determine cigs , holding other explanatory variables

fixed. That is rank condition holds and we can identify (6).

~~ln~~ cigprice and restaurn can be used as instruments for cigs . However, because $\ln \text{cigprice}$ is individually not statistically significant at any conventional significance level ($p\text{-value} = .951$), it cannot be used as instrument alone. (When we run reduced form regression but drop restaurn , $\ln \text{cigprice}$ is still ^{statistically} insignificant at any conventional level and dropping $\ln \text{cigprice}$ from reduced form equation ~~will~~ change the result only slightly.) Individually, restaurn can be used as IV for cigs while $\ln \text{cigprice}$ cannot, assuming that both are exogenous.

good!

(Incidence and return as IVs for cigs)

$$3.4) \ln \text{income} = 7.781 - 0.042 \text{cigs} + 0.0397 \text{educ} + 0.0938 \text{age} - 0.001 \text{age}^2$$

(.229) (.026) (.016) (.024) (.0003)

(Note that we get very similar result when using only return as IV for cigs)

$$3.5) \text{ OLS: } \ln \widehat{\text{income}} = 7.795 + 0.0017 \text{cigs} + 0.06 \text{educ} + 0.058 \text{age} - 0.0006 \text{age}^2$$

(.17) (.0017) (.0079) (.0076) (.000)

OLS, $\hat{\beta}_1^{\text{OLS}} = 0.0017$ means that if average number of cigarettes smoked per day increases by 1, income is expected to increase by 0.17%, on average, holding all other explanatory variable fixed. This implies that cigs has a positive effect on income which is quite not what we expect. However, we cannot reject $H_0: \beta_1 = 0$ at any conventional significant level (p-value = 0.313). Therefore, we can at best conclude that there is not enough evidence that number of cigarettes smoked per day has an ceteris paribus effect on income.

When we use IV estimator, $\hat{\beta}_1^{\text{IV}} = -0.042$ implying that, holding other factors constant if individual increases his/her cigarette consumption per day by 1 (on average), his/her income is expected to fall by average of 4.2%. Although this coefficient estimate is also statistically insignificant at 0.1 significant level (p-value = 0.107) it has expected sign.



Standard error of $\hat{\beta}_1$ (and also of other β) are much larger when we use IV than when we use OLS. This can be one of the reasons why our IV estimates for β_1 is still statistically insignificant even though it is much larger (in absolute term) when we compare it with OLS estimate.

Because OLS estimate for β_1 has unexpected sign, we suspect that we are facing endogeneity problem when using OLS and our OLS estimator suffers from asymptotic bias.

Question 2.1

Table1

. reg open lpcinc lland

Source	SS	df	MS	Number of obs = 114		
Model	28606.1936	2	14303.0968	F(2, 111) =	45.17	
Residual	35151.7966	111	316.682852	Prob > F =	0.0000	
				R-squared =	0.4487	
				Adj R-squared =	0.4387	
				Root MSE =	17.796	

open	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
lpcinc	.5464812	1.49324	0.37	0.715	-2.412473	3.505435
lland	-7.567103	.8142162	-9.29	0.000	-9.180527	-5.953679
_cons	117.0845	15.8483	7.39	0.000	85.68005	148.489

Table2

. ivregress 2sls inf (open=lland) lpcinc

Instrumental variables (2SLS) regression

Number of obs = 114
Wald chi2(2) = 5.73
Prob > chi2 = 0.0570
R-squared = 0.0309
Root MSE = 23.52

inf	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
open	-.3374871	.1422122	-2.37	0.018	-.6162179	-.0587563
lpcinc	.3758247	1.98839	0.19	0.850	-3.521348	4.272997
_cons	26.89934	15.1972	1.77	0.077	-2.886624	56.6853

Instrumented: open
Instruments: lpcinc lland

Question 2.2

Table3

	(1)	(2)	(3)
	inf	inf	inf
	b/se	b/se	b/se
open	-0.3375** (0.142)	-0.2150** (0.093)	-0.3329** (0.139)
lpcinc	0.3758 (1.988)		
_cons	26.8993* (15.197)	25.2342*** (4.102)	29.6066*** (5.608)
r2	0.031	0.045	0.032
N	114	114	114

* p<0.1, ** p<0.05, *** p<0.01

Question 2.3

Table 4

. ivregress 2sls inf (open=lland) lpcinc oil

Instrumental variables (2SLS) regression

Number of obs = 114
 Wald chi2(3) = 6.24
 Prob > chi2 = 0.1006
 R-squared = 0.0349
 Root MSE = 23.471

inf	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
open	-.3369707	.1419215	-2.37	0.018	-.6151316	-.0588097
lpcinc	.8032896	2.080441	0.39	0.699	-3.274299	4.880878
oil	-6.555731	9.627909	-0.68	0.496	-25.42609	12.31462
_cons	24.00886	15.75165	1.52	0.127	-6.863805	54.88153

Instrumented: open
 Instruments: lpcinc oil lland

Question 3.3

Table 5

. reg cigc educ age agesq lcigpric restaurn

Source	SS	df	MS	Number of obs = 807	
Model	7740.15214	5	1548.03043	F(5, 801) =	8.61
Residual	144013.531	801	179.792173	Prob > F =	0.0000
Total	151753.683	806	188.280003	R-squared =	0.0510
				Adj R-squared =	0.0451
				Root MSE =	13.409

cigc	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
educ	-.4501466	.1616396	-2.78	0.005	-.7674338	-.1328594
age	.822541	.1543224	5.33	0.000	.5196168	1.125465
agesq	-.0095903	.0016792	-5.71	0.000	-.0128864	-.0062942
lcigpric	-.3513161	5.76555	-0.06	0.951	-11.66869	10.96606
restaurn	-2.736389	1.109693	-2.47	0.014	-4.914639	-.5581394
_cons	1.580112	23.69558	0.07	0.947	-44.93266	48.09289

test lcigpric restaurn

- (1) lcigpric = 0
- (2) restaurn = 0

F(2, 801) = 3.13
 Prob > F = 0.0441

Question 3.4

Table 6

. ivregress 2sls lincome (cigs=lcigpric restaurn) educ age agesq

Instrumental variables (2SLS) regression

Number of obs = 807
 Wald chi2(4) = 89.80
 Prob > chi2 = 0.0000
 R-squared = .
 Root MSE = .87723

lincome	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
cigs	-.0421257	.0261371	-1.61	0.107	-.0933535	.009102
educ	.0396746	.0162305	2.44	0.015	.0078633	.0714859
age	.0938182	.0237794	3.95	0.000	.0472115	.1404249
agesq	-.0010508	.0002735	-3.84	0.000	-.0015868	-.0005148
_cons	7.780893	.2291541	33.95	0.000	7.33176	8.230027

Instrumented: cigs

Instruments: educ age agesq lcigpric restaurn

Question 3.5

Table 7

. reg lincome cigs educ age agesq

Source	SS	df	MS	Number of obs = 807		
Model	67.5412888	4	16.8853222	F(4, 802) =	39.61	
Residual	341.854549	802	.426252555	Prob > F =	0.0000	
Total	409.395838	806	.507935283	R-squared =	0.1650	
				Adj R-squared =	0.1608	
				Root MSE =	.65288	

lincome	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
cigs	.0017306	.0017137	1.01	0.313	-.0016333	.0050945
educ	.0603606	.0078983	7.64	0.000	.0448567	.0758645
age	.0576908	.0076436	7.55	0.000	.042687	.0726946
agesq	-.0006306	.0000834	-7.56	0.000	-.0007943	-.0004669
_cons	7.795444	.1704271	45.74	0.000	7.460908	8.129979

Table 8

	(1)	(2)	(3)
	lincome	lincome	lincome
	b/se	b/se	b/se
cigs	-0.0421 (0.02614)	-0.0414 (0.02595)	0.0017 (0.00171)
educ	0.0397** (0.01623)	0.0400** (0.01611)	0.0604*** (0.00790)
age	0.0938*** (0.02378)	0.0932*** (0.02361)	0.0577*** (0.00764)
agesq	-0.0011*** (0.00027)	-0.0010*** (0.00027)	-0.0006*** (0.00008)
_cons	7.7809*** (0.22915)	7.7811*** (0.22742)	7.7954*** (0.17043)
r2	.	.	0.165
N	807	807	807

* p<0.1, ** p<0.05, *** p<0.01