

Lecture 9 - Instrumental Variables Part II

1. Review of instrumental variables:

$$y_i = \alpha + \rho S_i + \varepsilon_i \quad (\text{structural eqn})$$

S_i is endogenous so instrument:

$$(a) \quad S_i = \pi_{10} + \pi_{11} Z_i + \eta_i \quad (\text{first stage})$$

$$y_i = \pi_{20} + \pi_{21} Z_i + u_i \quad (\text{reduced form})$$

Plugging in first-stage into structural:

$$y_i = \alpha + \rho(\pi_{10} + \pi_{11} Z_i + \eta_i) + \varepsilon_i$$

$$y_i = (\alpha + \rho\pi_{10}) + \rho\pi_{11} Z_i + (\rho\eta_i + \varepsilon_i)$$

$$y_i = \pi_{20} + \pi_{21} Z_i + u_i$$

$$\text{Taking } \frac{\pi_{21}}{\pi_{11}} = \frac{\text{reduced-form coeff } Z_i}{\text{first-stage coeff } Z_i} = \frac{\rho\pi_{11}}{\pi_{11}} = \rho$$

↓
causal estimate?

(b) In order for 2SLS estimate to be causal:

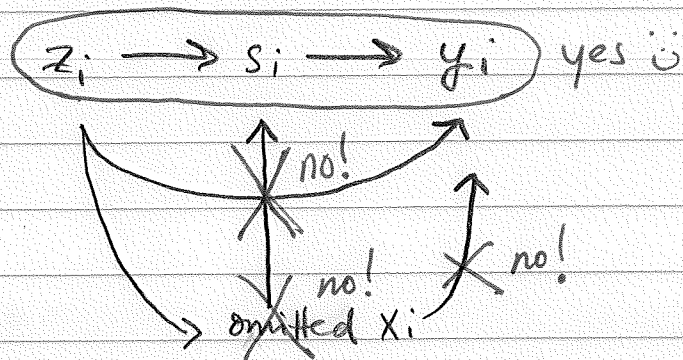
i) strong instrument:

$$S_i = \pi_{10} + \pi_{11} Z_i + \eta_i$$

↳ π_{11} is $\neq 0$, high R^2 (or F)

ii) Z_i is a "valid" instrument. \Rightarrow

z_i only affects y_i through its effect on s_i .



(c) when you run 2SLS with multiple instruments:

- i) F-stat ≥ 10 . The larger, the better.
- ii) Hansen-Sargen p-value $\geq .10$. The larger, the better (i.e., the closer to 1).

$\Leftrightarrow H_0$: instruments are not invalid

H_a : instruments are invalid

$E[z_i \hat{\varepsilon}_i] \neq 0$. In other words, instruments are correlated with error term

(If we "accept" H_0 , then we think instruments are valid)

"Accepting" H_0 means we cannot reject H_0 , which means p-value $\geq .10$.

So we really DO NOT want to reject H_0 in favor of H_a . Note that "accepting" H_0 does not mean instruments are valid. It means we think they are not invalid.

How to run Hansen-Sargan Test:

- estimate by 2SLS:

$$y_i = \alpha + \rho s_i + X_i + \varepsilon_i \text{ where}$$

$$s_i = \pi_{10} + \pi_{11} z_{i1} + \pi_{12} z_{i2} + X_i + \eta_i$$

- obtain $\hat{\varepsilon}_i$

- estimate by OLS

$$\hat{\varepsilon}_i = \gamma_{10} + \gamma_{11} z_{i1} + \gamma_{12} z_{i2} + X_i + \xi_i$$

- obtain R^2

$$\text{Test } \gamma_{10} = 0 \text{ and } \gamma_{12} = 0$$

$R^2 \approx 0$ if

z uncorrelated

with $\hat{\varepsilon}_i \Rightarrow$

instrument not
invalid.

$$NR^2 \sim \text{chi-square}(\#z_i\text{'s} - \#s_i\text{'s})$$

\rightarrow obtain p-value.

2. Wald Estimator : special case of IV
- when have 1 endog variable and 1 instrument
- AND - when z_i is binary: $z_i = \begin{cases} 1 \\ 0 \end{cases}$

$$Y_i = \alpha + \rho S_i + \varepsilon_i$$

$$S_i = \pi_{10} + \pi_{11} z_i + \eta_i$$

$$\rho = \frac{E[Y_i | z_i = 1] - E[Y_i | z_i = 0]}{E[S_i | z_i = 1] - E[S_i | z_i = 0]} = \frac{\text{reduced form est.}}{\text{1st stage est.}}$$

$$\Rightarrow E[Y_i | z_i] = \alpha + \rho E[S_i | z_i] + \underbrace{E[\varepsilon_i | z_i]}_{=0 \text{ if } \text{cov}(\varepsilon_i, z_i) = 0}$$

Example: Wald Estimate of Return to Schooling Using Quarter of Birth:

TABLE 4.1.2
Wald estimates of the returns to schooling using
quarter-of-birth instruments

	(1) Born in 1st Quarter of Year	(2) Born in 4th Quarter of Year	(3) Difference (Std. Error) (1) - (2)
ln (weekly wage)	5.892	5.905	-.0135 (.0034)
Years of education	12.688	12.839	-.151 (.016)
Wald estimate of return to education			.089 (.021)
OLS estimate of return to education			.070 (.0005)

Notes: From Angrist and Imbens (1995). The sample includes native-born men with positive earnings from the 1930-39 birth cohorts in the 1980 census 5 percent file. The sample size is 162,515.

$$Y_i = \pi_{20} + \pi_{21} z_i + u_i \quad (\text{reduced form})$$

$$z_i = \begin{cases} 1 & \text{if born in 1st quarter,} \\ 0 & \text{otherwise} \end{cases}$$

ln wages:

$$5.905 = \pi_{20} + \pi_{21}(0) \quad \text{if born in 4th Q}$$

$$\Rightarrow \pi_{20} = 5.905$$

$$5.892 = \pi_{20} + \pi_{21}(1) \quad \text{if born in 1st Q}$$

$$= 5.905 + \pi_{21}$$

$$\Rightarrow \pi_{21} = -0.0135 \Rightarrow \left\{ \begin{array}{l} \text{Being born in 1st Q} \\ \text{is associated with} \\ -0.0135 \text{ lower wages} \end{array} \right.$$

Years educ:

$$12.839 = \pi_{10} + \pi_{11}(0) \quad \text{if born in 4th Q}$$

$$\Rightarrow \pi_{10} = 12.839$$

$$12.688 = \pi_{10} + \pi_{11}(1) \quad \text{if born in 1st Q}$$

$$= 12.839 + \pi_{11}$$

$$\Rightarrow \pi_{11} = -0.151 \Rightarrow \left\{ \begin{array}{l} \text{Being born in 1st Q} \\ \text{is associated with} \\ -0.151 \text{ less years of} \\ \text{schooling.} \end{array} \right.$$

$$\frac{\frac{\Delta \text{wages}}{\Delta Q \text{ birth}}}{\frac{\Delta \text{schooling}}{\Delta Q \text{ birth}}} = \frac{\frac{-0.0135}{1}}{\frac{-0.151}{1}} = \frac{-0.0135}{-0.151} = 0.089 = \frac{\Delta \text{wages}}{\Delta \text{schooling}}$$

What's really nice about Wald Estimator?

- nice, intuitive, simple, can compute by hand! (Like we just did).

* - All 2SLS is just fancy versions of Wald.

i) if 1 endog and 2 dummy instrumental vars:

$$a) \quad Y_i = \alpha_1 + \rho_1 \hat{S}_{i1} + \varepsilon_i$$

where $\hat{S}_{i1} = \hat{\pi}_{10} + \hat{\pi}_{11} z_{i1}$

$$b) \quad Y_i = \alpha_2 + \rho_2 \hat{S}_{i2} + \varepsilon_i$$

where $\hat{S}_{i2} = \hat{\pi}_{20} + \hat{\pi}_{21} z_{i2}$

$$c) \quad Y_i = \alpha + \rho \hat{S}_i + \varepsilon_i$$

where $\hat{S}_i = \hat{\pi}_{30} + \hat{\pi}_{31} z_{i1} + \hat{\pi}_{32} z_{i2}$

$$\hat{\rho} \approx \frac{\hat{\rho}_1 + \hat{\rho}_2}{2} \approx (0.5) \hat{\rho}_1 + (0.5) \hat{\rho}_2$$

more generally:

$$\hat{\rho} = w \hat{\rho}_1 + (1-w) \hat{\rho}_2$$

$w \in (0,1)$ with w based on variance of $\hat{\varepsilon}_i$ from (a) and $\hat{\varepsilon}_i$ from (b).

$$w = \frac{\frac{n_a}{\sigma_{\varepsilon a}^2}}{\frac{n_a}{\sigma_{\varepsilon a}^2} + \frac{n_b}{\sigma_{\varepsilon b}^2}}$$

ii) Even if z_i is not a dummy $\{0, 1\}$ variable, it can often be made into a dummy.

e.g. z_i is $\{1, 2, 3, 4\}$ quarter of birth
gen four quarter-of-birth dummies:

\Rightarrow	z_1	z_2	z_3	z_4
Q1	1	0	0	0
Q2	0	1	0	0
Q3	0	0	1	0
Q4	0	0	0	1

iii) If we're thinking of monitoring and evaluation of "randomized" experiments,
 $z_i = \{0, 1\}$ is assignment to treatment.

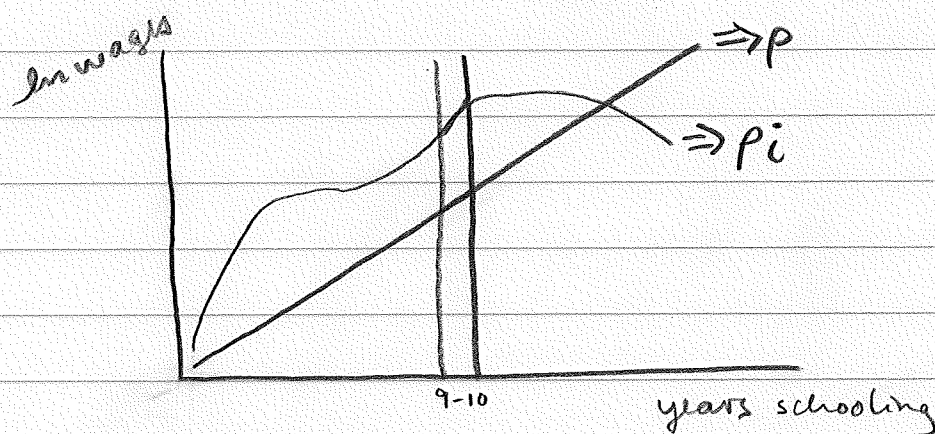
3. Interpreting IV Estimates.

$$Y_i = \alpha + \rho S_i + \varepsilon_i$$

or

$$Y_i = \alpha + \underline{\underline{\rho_i}} S_i + \varepsilon_i$$

ρ or ρ_i ? Homogenous or heterogeneous effects of schooling on wages.



Imagine a policy that compels individuals to remain in school until they're 16. Those that turn 16 earlier drop out at grade 10; those that turn 16 later drop out at grade 11.

From this instrument, $\hat{\rho} = .089 \Rightarrow$
1 extra year = 9% higher earnings.

- Generalizable?

- If we had induced one extra year between 8th - 9th grade, would $p = .089$?

- Average Population?

- What kinds of people are induced to get one extra year of schooling but would have dropped out otherwise?

Who does instrument target?

never takers : would have dropped out much earlier anyways (even though illegal)

always takers : would have gotten much more school anyways.

compliers : induced into getting more school because of treatment.

LATE (Local Average Treatment Effect)

↳ Effect of 1 extra year of schooling between 10th - 11th grade for those who are induced into remaining in school because of minimum age laws is $p = .089$.