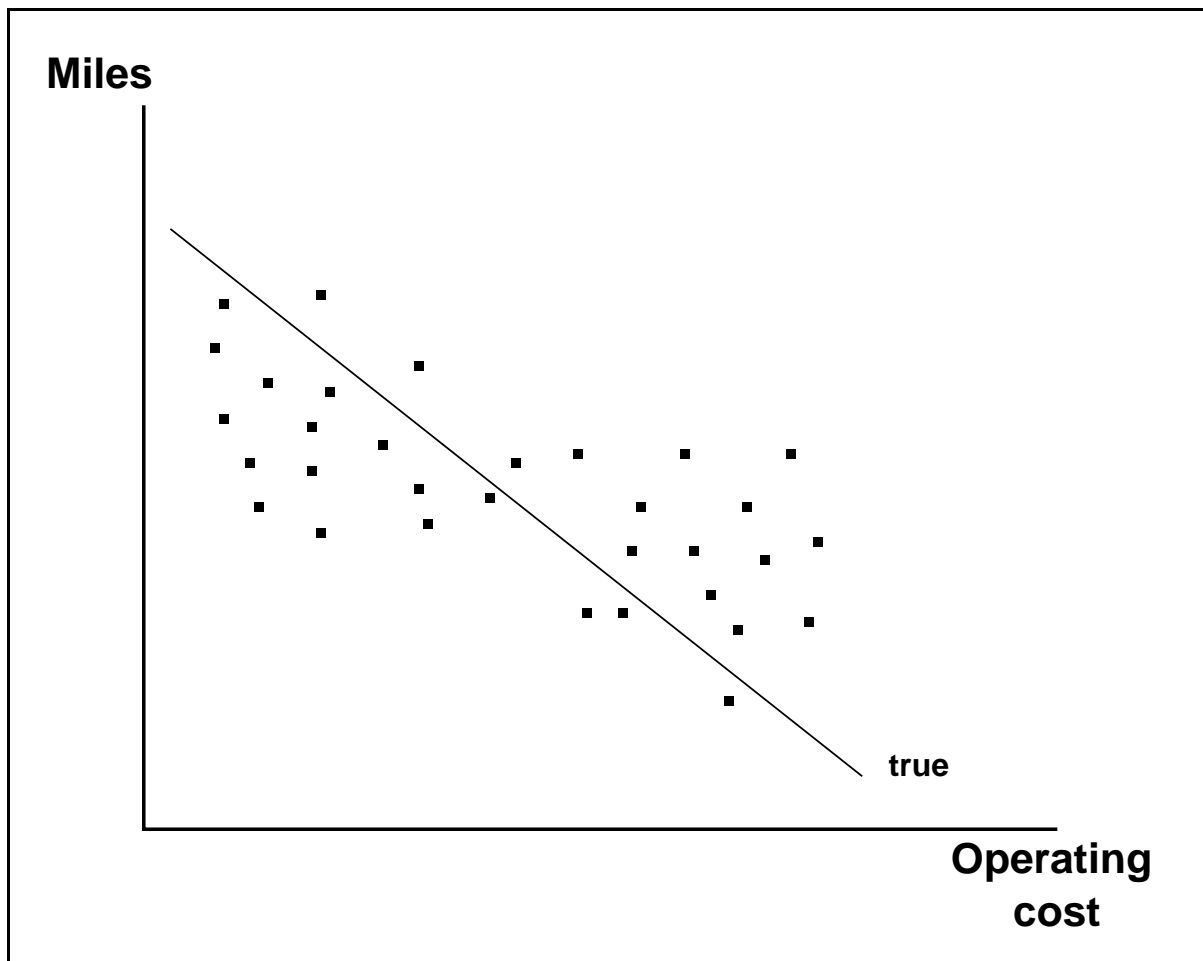


# LECTURE / DISCUSSION

## Instrumental Variables

# Instrumental Variables Estimation

Problem: error is correlated with explanatory variable.



$$\text{Miles} = \alpha + \beta(\text{op. cost}) + \varepsilon$$

**Solution #1:** Include the variable that causes the correlation, if possible.

$$(\text{Miles}) = \alpha + \beta(\text{op. cost}) + \theta(\text{distance to work}) + \varepsilon$$

**Solution #2: Instrumental variables estimation**

$$Y_n = \alpha + \beta X_n + \varepsilon_n$$

$$\text{Corr}(X, \varepsilon) \neq 0$$

Find a variable,  $Z$ , that is

- correlated with  $X$
- not correlated with  $\varepsilon$

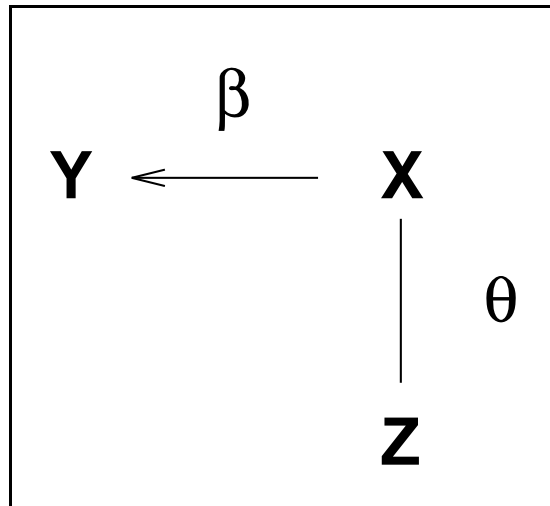
$Z$  is called an **instrument**.

$$\hat{\beta}_{IV} = \frac{\sum z_n y_n}{\sum z_n x_n} = \frac{\text{cov}(z, y)}{\text{cov}(z, x)}$$

**Example:** Miles =  $\alpha + \beta(\text{op. cost}) + \varepsilon$   
Instrument: gas price

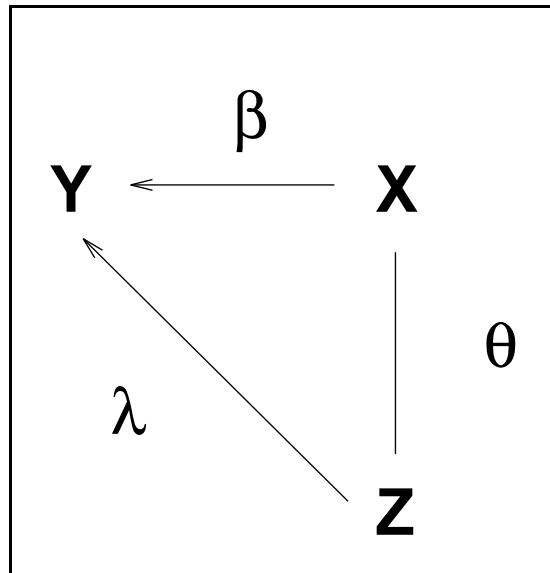
$$\hat{\beta}_{IV} = \frac{\text{cov}(\text{gas price}, \text{miles})}{\text{cov}(\text{gas price}, \text{op. cost})}$$

## Why Does IV Work?



$$E(\hat{\beta}_{IV}) = \frac{\text{cov}(zy)}{\text{cov}(zx)} = \frac{\beta\theta}{\theta} = \beta$$

## When IV is Biased



$$E(\hat{\beta}_{IV}) = \frac{\text{cov}(zy)}{\text{cov}(zx)} = \frac{\theta\beta + \lambda}{\theta} = \beta + \frac{\lambda}{\theta}$$

## Why Is IV Unbiased?

$$y_n = \beta x_n + \varepsilon_n$$

$$\sum z_n y_n = \beta \sum z_n x_n + \sum z_n \varepsilon_n$$

$$\begin{aligned} E(\sum z_n y_n) &= \beta E(\sum z_n x_n) + E(\sum z_n \varepsilon_n) \\ &= \beta E(\sum z_n x_n) \end{aligned}$$

So that  $z_n$  isolates the covariation between  $y_n$  and  $x_n$ .  
This leads to

$$\hat{\beta}_{IV} = \frac{\sum z_n y_n}{\sum z_n x_n}$$

- ➡ Efficiency is greater with instruments that are more highly correlated with  $X$ , while still uncorrelated with the error terms.

## Can Apply IV Through 2SLS

$$Y_n = \alpha + \beta X_n + \varepsilon_n$$

instrument  $Z_n$

Step 1: Run OLS regression

$$X_n = \lambda + \theta Z_n + \mu_n$$

Get predicted  $\hat{X}_n = \hat{\lambda} + \hat{\theta} Z_n$

Step 2: Run OLS regression using  $\hat{X}$

$$Y_n = \alpha + \beta \hat{X}_n + \varepsilon_n^*$$

## Example

$$\text{Miles} = \alpha + \beta(\text{op. cost}) + \varepsilon$$

instrument: gas price

Step 1: Regress

$$(\text{op. cost}) = \lambda + \theta(\text{gas price}) + \mu$$

Calculate

$$(\text{op. cost}) = \hat{\lambda} + \hat{\theta}(\text{gas price})$$

Step 2: Regress

$$\text{Miles} = \alpha + \beta(\text{op. cost}) + \varepsilon^*$$

## 2SLS Interpretation of IV

$$Y_n = \alpha + \beta X_n + \varepsilon_n$$

Step 1:  $X$  is decomposed into a part that is uncorrelated with  $\varepsilon$  and a part that is correlated with  $\varepsilon$ .

$$X_n = \lambda + \theta Z_n + \mu_n$$


uncorrelated correlated  
with  $\varepsilon$  with  $\varepsilon$

Step 2: Use the portion of  $X$  that is uncorrelated with  $\varepsilon$  in estimating the original equation.

$$Y_n = \alpha + \beta \hat{X}_n + \varepsilon_n^*$$

$$\hat{X}_n = \hat{\lambda} + \hat{\theta} z_n$$

$$\varepsilon_n^* = \varepsilon_n + \beta \hat{\mu}_n$$

The part of  $X$  that was correlated with  $\varepsilon$  gets moved into the error.

The part of  $X$  that was uncorrelated with  $\varepsilon$  stays as an explanatory variable.

2SLS method also shows that many instruments can be used in IV.

$$(\text{Miles}) = \alpha + \beta(\text{op. cost}) + \varepsilon$$

Instruments:    gas price  
                    relative prices for large and small cars

Not instruments:    income  
                          family size

Note: Income and family size **can** serve as instruments if they also enter the original regression as explanatory variables.

$$\text{Miles} = \alpha + \beta(\text{op. cost}) + \theta(\text{income}) + \lambda(\text{family size}) + \varepsilon$$

Instruments:    gas price  
                    relative prices for large and small cars  
                    income  
                    family size

Because income and family size enter directly, they are not part of  $\varepsilon$  and hence are not correlated with  $\varepsilon$ .

## Equivalence of IV and 2SLS

2SLS

Step 1:

$$x_n = \theta z_n + \varepsilon_n$$

$$\hat{\theta}_{OLS} = \frac{\text{cov}(zx)}{\text{var}(z)}$$

$$\hat{x}_n = \hat{\theta}_{OLS} z_n$$

Step 2:

$$y_n = \beta \hat{x}_n + \varepsilon_n$$

$$\hat{\beta}_{2SLS} = \frac{\text{cov}(\hat{x}y)}{\text{var}(\hat{x})}$$

$$\begin{aligned} \hat{\beta}_{2SLS} &= \frac{\text{cov}(\hat{\theta}_{OLS}zy)}{\text{var}(\hat{\theta}_{OLS}z)} = \frac{\hat{\theta}_{OLS} \text{cov}(zy)}{\hat{\theta}_{OLS}^2 \text{var}(z)} \\ &= \frac{\text{cov}(zy)}{\hat{\theta}_{OLS} \text{var}(z)} = \frac{\text{cov}(zy)}{\frac{\text{cov}(zx)}{\text{var}(z)} \cdot \text{var}(z)} \\ &= \frac{\text{cov}(zy)}{\text{cov}(zx)} = \hat{\beta}_{IV} \end{aligned}$$

# TSP Commands

Three ways to estimate

$$Y = \alpha + \beta X + \varepsilon \quad \text{with instrument } Z$$

## 1. 2SLS explicitly as two OLS regressions

```
olsq x c,z;  
genr xhat = @fit;  
olsq y c,xhat;
```

## 2. 2SLS implicitly

```
2sls(inst = (c,z)) y c,x;
```

## 3. IV

```
inst y c,x invr c,z;
```