

Instrumental Variables

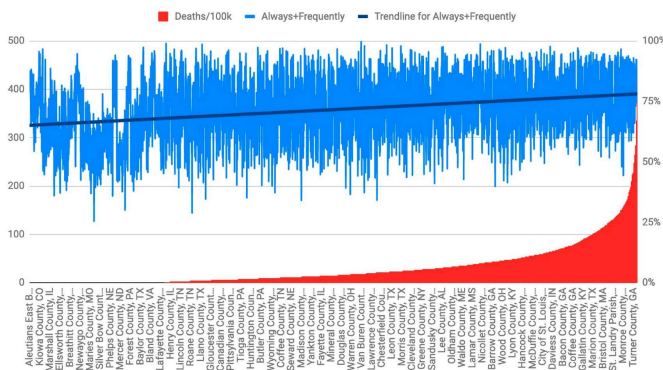
Monday, November 13, 2023 1:58 PM

From Khoa Vu, University of Minnesota:



From some guy on Twitter:

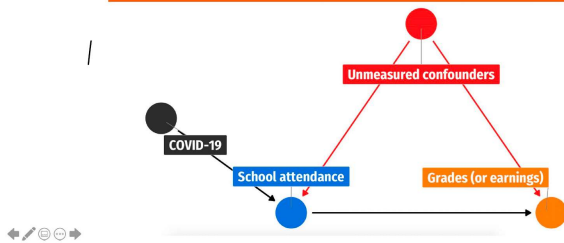
Mask Use "Always+Frequently" vs Deaths/100k [by US County]



From Andrew Heiss, Georgia State University:

COVID-19 as an instrument

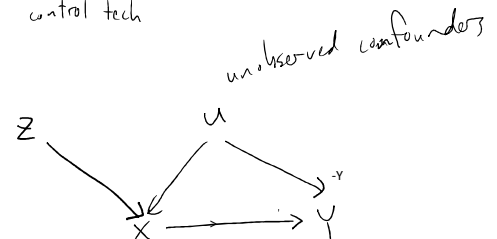
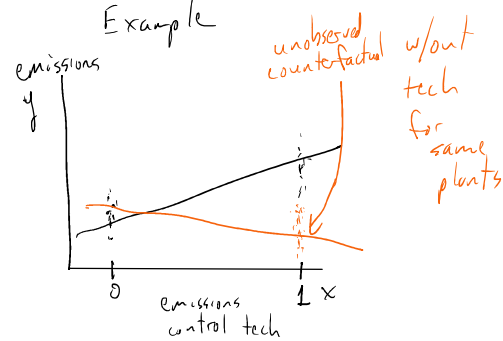
What effect does closing schools have on student performance or lifetime earnings?



$$y_t = \beta_0 + \beta_1 x_t + \varepsilon_t$$

$$E(x_t \cdot \varepsilon_t) = 0$$

Example

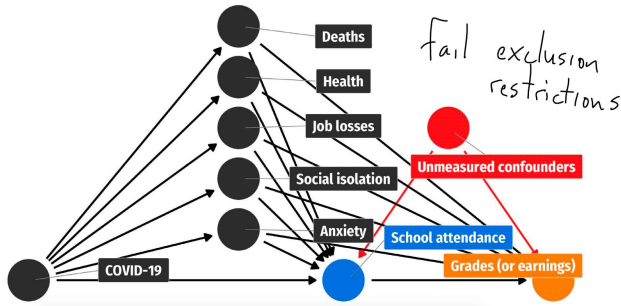


Z affects X, not U
and only affects Y through X

Z an Instrumental Variable "IV"

① Relevance : $Cov(Z, X) \neq 0$

lolnope



(2) Exclusion Restriction : $Cov(Z, u) = 0$

In commodity markets \rightarrow want to know shape of supply + demand curves

$$q_{st} = \alpha + \beta P_t + u_t$$

$$q_{dt} = c - d P_t + v_t$$

In practice we observe equilibrium outcomes

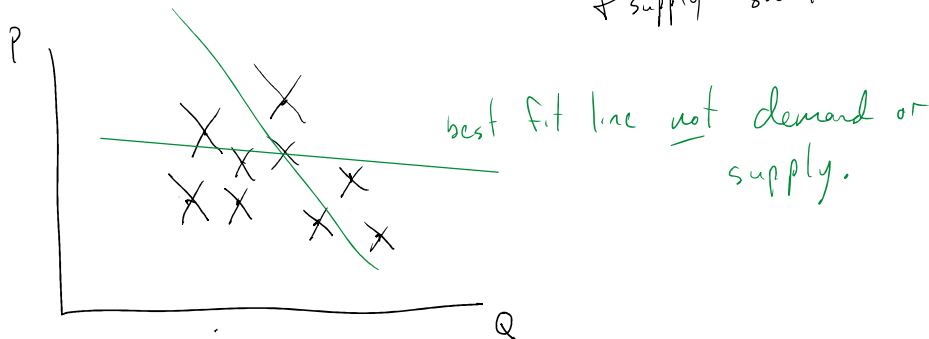
$$\alpha + \beta P_t + u_t = c - d P_t + v_t$$

$$\text{solve for } P_t = \frac{c - \alpha}{\beta + d} + \frac{v_t + u_t}{\beta + d}$$

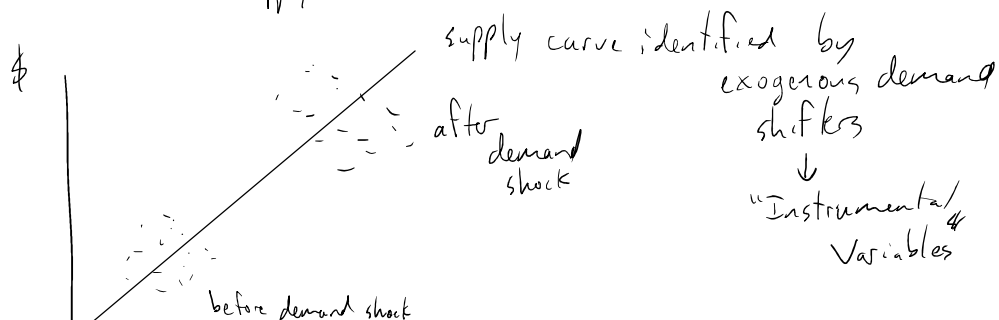
Naive Regression :

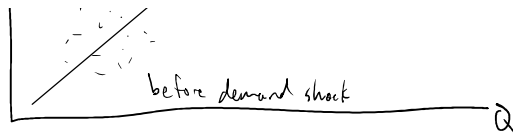
$$q_t = \phi_0 + \phi_1 (P_t) + \varepsilon_t$$

contains $v_t + u_t$ (pointing to ε_t)
correlated \rightarrow both contain unobserved demand + supply shocks



Maybe I can find some demand shifters to trace out supply curve





Variables

Basic Idea :

Endogenous Variable X , Exogenous (Instrumental Variable) Z

1) regress X on Z , predict \hat{X} from that regression

$$X_t = \underbrace{\phi_0 + \phi_1 z_t}_{\text{exogenous part of } X} + \underbrace{v_t}_{\text{endogenous part of } X}$$

$$\hat{X}_t = \hat{\phi}_0 + \hat{\phi}_1 z_t$$

→ contains variation only from $z \rightarrow$ shared variation between x + z

2) regress Y on \hat{X}

$$y_t = \beta_0 + \beta_1 \hat{x}_t + e_t$$

$$y_t = \hat{\beta}_0 + \hat{\beta}_1 \hat{x}_t + \hat{e}_t \quad \text{fitted regression}$$

Back to supply + demand

$$q_{st} = \beta_0 + \beta_1 p_t + \beta_2 z_{st} + u_t \quad z_{st} = \text{input prices, technology shocks, war, weather shocks}$$

$$q_{dt} = \alpha_0 - \alpha_1 p_t + \alpha_2 z_{dt} + v_t \quad z_{dt} = \text{consumer income}$$

suppose only P_t is endogenous in each regression

$$E(z_{st} u_t) = 0 \quad E(z_{dt} v_t) = 0$$

instruments for demand: z_{dt} for itself, z_{st} for P_t

" " supply: z_{st} for itself, z_{dt} for P_t

Focus on demand for now:

$$q_{dt} = \underline{x}_t^T \cdot \underline{\alpha} + v_t = \begin{bmatrix} 1 & p_t & z_{dt} \end{bmatrix} \begin{bmatrix} \alpha_0 \\ -\alpha_1 \\ \alpha_2 \end{bmatrix} + v_t$$

every row ↗

$$\underline{q_d} = \underline{X} \underline{\alpha} + \underline{v}$$

$T \times 1 \quad T \times 3 \quad 3 \times 1 \quad T \times 1$

$$\wedge \quad 1 \dots T \quad 1 \dots T \quad - \text{matrix} \quad \underline{\text{Cov}(q_d, x)}$$

$$\begin{matrix} U \\ T \times 1 \end{matrix} \quad \begin{matrix} T \times 3 & 3 \times 1 \\ T \times 1 \end{matrix}$$

From OLS $\hat{\alpha} = \left(\underset{3 \times T}{\tilde{X}^T \tilde{X}} \right)^{-1} \underset{3 \times T}{\tilde{X}^T} \underset{3 \times 1}{q_d} = \text{matrix version of } \frac{\text{Cov}(q_d, x)}{\text{Var}(x)}$

For IV (or 2-stage least squares 2SLS)

1) regress \tilde{x} on \tilde{z} : $\left(\underset{\sim}{\tilde{z}^T \tilde{z}} \right)^{-1} \underset{\sim}{\tilde{z}^T} \tilde{x}$ coefficients in first stage

predict $\hat{\tilde{x}} = \underset{\sim}{\tilde{z}} \left[\left(\underset{\sim}{\tilde{z}^T \tilde{z}} \right)^{-1} \underset{\sim}{\tilde{z}^T} \tilde{x} \right]$
 right hand side of stage 1 estimated coefficients in stage 1

2) regress q_d on $\hat{\tilde{x}}$
 $\hat{\alpha}_{IV} = \left(\underset{\sim}{\hat{\tilde{x}}^T \hat{\tilde{x}}} \right)^{-1} \underset{\sim}{\hat{\tilde{x}}^T} \cdot q_d$

$$= \left[\tilde{x}^T \tilde{z} \left(\tilde{z}^T \tilde{z} \right)^{-1} \tilde{z}^T \tilde{x} \right]^{-1} \tilde{x}^T \tilde{z} \left(\tilde{z}^T \tilde{z} \right)^{-1} \tilde{z}^T q_d$$

$$= \left[\tilde{x}^T \tilde{z} \left(\tilde{z}^T \tilde{z} \right)^{-1} \tilde{z}^T \tilde{x} \right]^{-1} \tilde{x}^T \tilde{z} \left(\tilde{z}^T \tilde{z} \right)^{-1} \tilde{z}^T q_d$$

general case w/ more IV's (supply shifter) than endogenous variables (price)

In special case w/ 1 IV for each endogenous variable, formula simplification:

$$\hat{\alpha}_{IV} = \left(\underset{\sim}{\tilde{z}^T \tilde{x}} \right)^{-1} \underset{\sim}{\tilde{z}^T} q_d \quad \text{matrix version of } \frac{\text{Cov}(\tilde{z}, y)}{\text{Cov}(\tilde{z}, x)}$$

use these formulas to get standard errors:

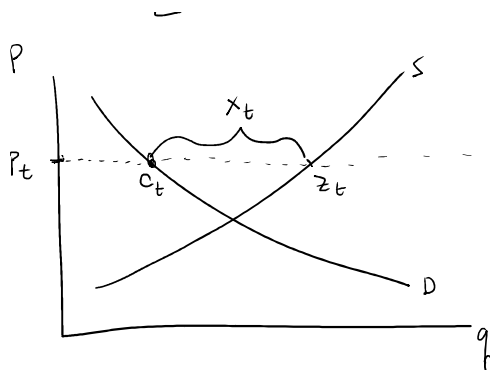
$$\hat{\alpha}_{IV} = \left(\underset{\sim}{\tilde{z}^T \tilde{x}} \right)^{-1} \underset{\sim}{\tilde{z}^T} (\underset{\sim}{x} \alpha + v)$$

Examples from commodity markets:

storable commodity $\left\{ \begin{array}{l} \text{Roberts + Schlenker American Economic Review 2013} \\ \quad \rightarrow \text{ethanol} \\ \text{Hausman + Kellogg Brookings Economic Papers 2015} \\ \quad \rightarrow \text{shale gas} \end{array} \right.$

P 1, x, / S

x_t = amount stored in t carried into $t+1$ as inventory



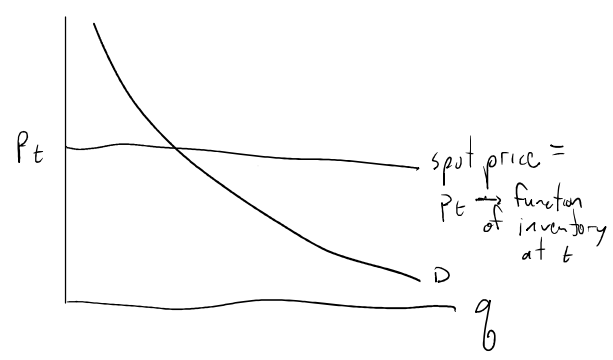
x_t = amount stored in t
carried into $t+1$ as inventory

$$z_{t+1} = x_t + f(\lambda_t, \omega_{t+1})$$

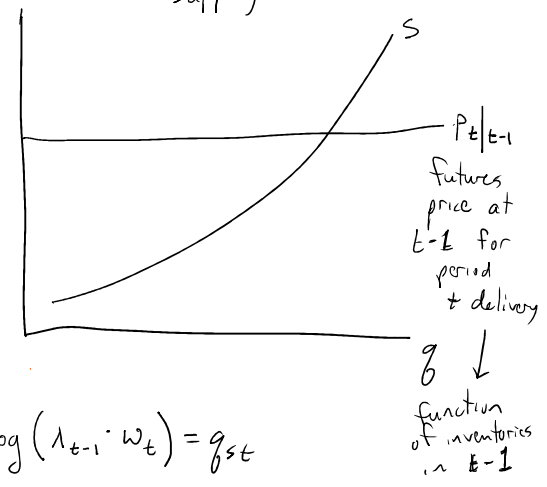
↑
storage
decisions
previously
↑
inputs
decisions
chosen
previously
↑
production
function
productivity
shock
unforecastable

insight: if we can capture incentives for storage in earlier periods, we can trace out supply and demand.

Demand



Supply



New production at time $t = \log(\lambda_{t-1} \cdot \omega_t) = q_{st}$

as a supply curve:
2nd stage

$$q_{st} = \alpha_s + \beta_s P_{t|t-1} + \gamma_s \omega_{st} + \underbrace{f_{sz}(t)}_{\text{monthly dummies, trend}} + u_t$$

first stage of IV for $P_{t|t-1}$

$$P_{t|t-1} = \delta_s + \mu_{s0} \cdot \omega_{st} + \underbrace{\mu_{s1} \omega_{st-1} + \mu_{s2} \omega_{dt-1}}_{\text{correlated}} + \underbrace{f_{s1}(t)}_{\text{monthly dummies, trends}} + \varepsilon_t$$

can't be correlated

good IV has two requirements:

- ① IV's strongly correlated w/ endogenous x 's
- ② NOT correlated with $u_t \rightarrow$ the main equation residual

Demand side:

$$q_{dt} = \alpha_d + \beta_d P_t + \omega_{dt} + f_{dz}(t) + v_t$$

↑
spot price

first stage

$$P_t = \delta_1 + \mu_{10} \cdot \omega_{dt} + f_{d1}(t) + \underbrace{\mu_{d1} \omega_{dt-1} + \mu_{d2} \omega_{st-1}}_{\text{correlated}} + e_t$$

first stage

$$P_t = \delta_d + \mu_{do} \cdot \omega_{dt} + f_{dl}(t) + \underbrace{\mu_{dl} \omega_{dt-1} + \mu_s \omega_{st-1}}_{\text{Instruments}} + e_t$$

What to use? for $\omega_{dt-1}, \omega_{st-1}$

Roberts + Schlenker (ethanol) unexpected corn yield shocks
 → regress yield on time trends, seasonal dummies,
 take residuals → ω_{st}

Hausman + Kellogg (natural gas)

→ neighboring states demand shocks (weather in neighboring states)
 → neighbor's demand shock is a supply shock for me.

→ lagged cumulative weather shock in neighboring states over a year
 → want shocks to inventory!

Cournot Econometrics (Vitamins Paper)

Cournot Problem: small # firms, each influences market price

Competitive market each firm "i"

$$\max_{q_i} \pi(q_i) = p \cdot q_i - c_i(q_i)$$

FOC: $p - c'_i(q_i) = 0$

Oligopoly w/ quantity competition

$$\begin{aligned} \max_{q_i} \pi_i(q_i) &= \underbrace{P(q_1 + \dots + q_i + \dots + q_N)}_{\text{inverse demand function}} \cdot q_i - c_i(q_i) \\ &= P(Q) \cdot q_i - c_i(q_i) \quad Q = \sum_{i=1}^N q_i \end{aligned}$$

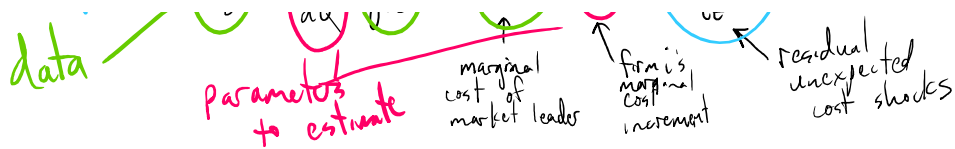
$$\text{FOC: } \underbrace{P(Q) + \frac{dP}{dQ} \cdot q_i}_{\text{marginal revenue}} - \underbrace{c'_i(q_i)}_{\text{marginal cost}} = 0$$

In Vitamins Paper: marginal costs are constant

Each firm!

$$\text{data } P_t + \underbrace{\frac{dP}{dQ}}_{\text{parameter } \delta} q_{it} = \underbrace{C_{mt}}_{\text{marginal cost, of } n} + \underbrace{\delta_i}_{\text{firm's marginal cost}} + \underbrace{\epsilon_{it}}_{\text{residual unexpected cost shocks}}$$

Regression residual



- approach: use C_{mt} as IV / supply shifter
- ① in demand function to get $\frac{dP}{dQ}$ demand slope.
 - ② use demand shifters as IV's to estimate the γ_i (the marginal cost functions)
 → disposable income in high income countries