

# Structural Models

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## Abstract

## 1. Causality in Structural Models

### 1.1. General Setup

- Say that after solving a model comprising a system of equations (eg supply and Demand functions) you end up with an equation for  $Y$ , say  $Y = F(x_1, x_2, \dots, x_n)$
- Now we are interested in the **Causal effect**, the response of  $Y$  to a *ceteris paribus* change in one variable. If we can vary the "cause"  $x_j$  independently, then:

$$\frac{\partial Y}{\partial x_j} = F'_j(x_1, x_2, \dots, x_n)$$

- This is the marginal causal effect of  $x_j$  on  $Y$ .
- If we assume linear equation:  $Y = \beta X + \varepsilon$ , we can use OLS to recover the parameter of interest.
- Under proper identification,  $\beta_j = \frac{\partial E[Y]}{\partial x_j}$  can be interpreted causally and  $\beta_j$  is called a structural parameter as it is derived from a structural model.

### 1.2. Interrelated Causes and Endogeneity

- When causes are interrelated, causal effect still exists but is harder to identify.
- Example: Demand and Supply

$$Y^d = Y^d(P^d, Z^d, U^d)$$

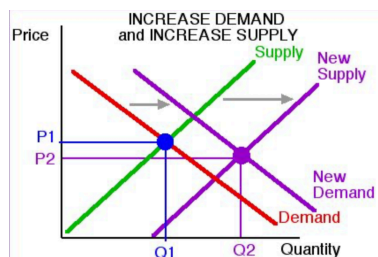
$$Y^s = Y^s(P^s, Z^s, U^s)$$

- If  $P^d$  and  $P^s$  could be varied exogenously:

$$\frac{\partial Y^d}{\partial P^d}, \quad \frac{\partial Y^s}{\partial P^s}$$

would be causal effects.

- But in equilibrium:  $P^d = P^s = P$ ,  $Y^d = Y^s = Y$ . So you do not specifically observe the two equations above!
- You are now observing an equilibrium point / a series of equilibrium points (say  $(Y, P)$ ) that is JOINTLY determined by S and D,
- The relation between  $y$  and  $p$  that you are observing is totally uninformative as you have 2 unobservables (D and S) determining just one observable: price may be determined by a simultaneous shift in D and supply.



- So causal interpretation of  $\frac{\partial Y}{\partial P}$  is invalid due to endogeneity (simultaneity bias).
- You want to isolate shifts in D or S

### 1.3. Using Exogenous Variation: Reduced Form

- If we can vary  $Z^d$  and/or  $Z^s$  exogenously, we can recover causal effects (IV style).
- Reduced form (endogenous as a function of exogenous variables):

$$Y = Y(Z^d, U^d, Z^s, U^s)$$

$$P = P(Z^d, U^d, Z^s, U^s)$$

- If  $Z^s$  affects  $P$  and can be exogenously varied (no effect on  $Z^d, U^d, U^s$ ), we can identify causal effect of  $Z^s$  on  $P$  and  $Y$ .
- idea is shifting the supply curve while keeping the demand curve fixed
- Reduced Form Causal Effects of  $Z^s$  on  $P$  and  $Y$  (ceteris paribus):

$$\frac{\partial Y}{\partial Z^s}, \quad \frac{\partial P}{\partial Z^s}$$

## 2. Structural Model: Concrete Example

### 2.1. Model Setup

- Structural equations:

$$Y^d = \beta_0 + \beta_1 P^d + U$$

$$Y^s = P^s - Z$$

- In equilibrium:  $P^d = P^s = P$ ,  $Y^d = Y^s = Y$

- Structural form:

$$Y = \beta_0 + \beta_1 P + U$$

$$P = Y + Z$$

we cannot identify the structural coefficient  $\beta_1$  representing the ceteris paribus marginal effect of  $P^d$  on  $Y^d$

### 2.2. Solving the System

- Assumption:  $U$  and  $Z$  are independent!
- Solve the system, reduced form of the model:

$$Y = \frac{\beta_0}{1 - \beta_1} + \frac{\beta_1}{1 - \beta_1} Z + \frac{U}{1 - \beta_1}$$

$$P = \frac{\beta_0}{1 - \beta_1} + \frac{1}{1 - \beta_1} Z + \frac{U}{1 - \beta_1}$$

- Let  $\gamma_0 = \frac{\beta_0}{1 - \beta_1}$ ,  $\gamma_1 = \frac{\beta_1}{1 - \beta_1}$ ,  $\gamma_2 = \frac{1}{1 - \beta_1}$
- Reduced form causal effects:

$$\frac{\partial Y}{\partial Z^s} = \gamma_1, \quad \frac{\partial P}{\partial Z^s} = \gamma_2$$

- Recover structural parameter of interest from the reduced parameters:

$$\frac{\partial Y^d}{\partial P^d} = \beta_1 = \frac{\partial Y / \partial Z^s}{\partial P / \partial Z^s}$$

The Exogenous variation in Supply allowed you to find the slope of Demand

### Interpretation

- $\gamma_1 = \frac{\beta_1}{1 - \beta_1}$  is causal effect of  $Z^s$  on  $Y$ .

- Reduced form coefficient is the composition of two effects:

$$\frac{\partial Y}{\partial Z} = \frac{\partial Y^s}{\partial P^s} \cdot \frac{\partial P}{\partial Z} = \beta_1 \cdot \frac{1}{1 - \beta_1}.$$

Z is the exogenous variable generating the change in Y, while P is the intermediate cause.

### 3. Structural Models: Other Examples

#### 3.1. Identifying the Supply Slope

- New model:

$$Y^d = \beta_0 + \beta_1 P^d + Z + U$$

$$Y^s = \beta_2 P^s$$

- Goal: identify  $\beta_2$
- You can identify the slope of supply BECAUSE you have exogenous shocks in Demand!

### 4. Structural Approach and Potential Outcomes

- Assumptions needed to identify  $\beta$  in a structural equation are equivalent to potential outcome assumptions.
- Structural equation  $Y_i = \beta D_i + \varepsilon_i$ 
  - Assumes homogeneous  $\beta$  (guarantees no treatment effect heterogeneity)
  - Selection bias arises if  $D_i$  is correlated with  $\varepsilon_i$  (some unobserved characteristics are correlated with the treatment assignment)