

# Interpreting IV, Part 1

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Extract from: Building Bridges Between Structural  
and Program Evaluation Approaches to Evaluating Policy  
James Heckman (JEL 2010)

Econ 312, Spring 2021

- Roy (1951): Agents face two potential outcomes  $(Y_0, Y_1)$  with distribution  $F_{Y_0, Y_1}(y_0, y_1)$  where “0” refers to a no treatment state and “1” refers to the treated state and  $(y_0, y_1)$  are particular values of random variables  $(Y_0, Y_1)$ .
- More generally, set of potential outcomes is  $\{Y_s\}_{s \in \mathcal{S}}$  where  $\mathcal{S}$  is the set of indices of potential outcomes.
- Roy model  $\mathcal{S} = \{0, 1\}$ .

- Analysts observe either  $Y_0$  or  $Y_1$ , but not both, for any person.
- In the program evaluation literature, this is called the **evaluation problem**.

- The **selection problem**.
- Values of  $Y_0$  or  $Y_1$  that are observed are not necessarily a random sample of the potential  $Y_0$  or  $Y_1$  distributions.
- In the original Roy model, an agent selects into sector 1 if  $Y_1 > Y_0$ .

$$D = \mathbf{1}(Y_1 > Y_0), \quad (1)$$

- Generalized Roy model ( $C$  is the cost of going from “0” to “1”)

$$D = \mathbf{1}(Y_1 - Y_0 - C > 0). \quad (2)$$

- The outcome observed for any person,  $Y$ , can be written as

$$Y = DY_1 + (1 - D)Y_0. \quad (3)$$

- $\mathcal{I}$  denotes agent information set.
- In advance of participation, the agent may be uncertain about all components of  $(Y_0, Y_1, C)$ .
- Expected benefit:  $I_D = E(Y_1 - Y_0 - C \mid \mathcal{I})$ .
- Then

$$D = \mathbf{1}(I_D > 0). \quad (4)$$

- The decision maker selecting “treatment” may be different than the person who experiences the outcomes ( $Y_0, Y_1$ ).

- The *ex-post* objective outcomes are  $(Y_0, Y_1)$ .
- The *ex-ante* outcomes are  $E(Y_0 | \mathcal{I})$  and  $E(Y_1 | \mathcal{I})$ .
- The *ex-ante* subjective evaluation is  $I_D$ .
- The *ex-post* subjective evaluation is  $Y_1 - Y_0 - C$ .
- Agents may regret their choices because realizations may differ from anticipations.



- $Y_1 - Y_0$  is the individual level treatment effect.
- Also, the Marshallian ceteris paribus causal effect.
- Because of the evaluation problem, it is generally impossible to identify individual level treatment effects.
- Even if it were possible,  $Y_1 - Y_0$  does not reveal the *ex-ante* subjective evaluation  $I_D$  or the *ex-post* assessment  $Y_1 - Y_0 - C$ .

- Economic policies can operate through changing ( $Y_0, Y_1$ ) or through changing  $C$ .

## Population Parameters of Interest

- Conventional parameters include the Average Treatment Effect ( $ATE = E(Y_1 - Y_0)$ ), the effect of Treatment on The Treated ( $TT = E(Y_1 - Y_0 | D = 1)$ ), or the effect of Treatment on the Untreated ( $TUT = E(Y_1 - Y_0 | D = 0)$ ).

- In positive political economy, the fraction of the population that perceives a benefit from treatment is of interest and is called the **voting criterion** and is

$$\Pr(I_D > 0) = \Pr(E(Y_1 - Y_0 - C \mid \mathcal{I}) > 0).$$

- In measuring support for a policy in place, the percentage of the population that *ex-post* perceives a benefit is also of interest:  $\Pr(Y_1 - Y_0 - C > 0)$ .

- Determining marginal returns to a policy is a central goal of economic analysis.
- In the generalized Roy model, the margin is specified by people who are indifferent between “1” and “0”, i.e., those for whom  $I_D = 0$ .
- The mean effect of treatment for those at the margin of indifference is

$$E(Y_1 - Y_0 \mid I_D = 0).$$

## Treatment Effects Versus Policy Effects

- Policy Relevant Treatment Effect (Heckman and Vytlacil, 2001) extends the Average Treatment Effect by accounting for voluntary participation in programs.
- “ $b$ ”: baseline policy (“before”) and “ $a$ ” represent a policy being evaluated (“after”).
- $Y^a$ : outcome under policy  $a$ ;  $Y^b$  is the outcome under the baseline.
- $(Y_0^a, Y_1^a, C^a)$  and  $(Y_0^b, Y_1^b, C^b)$  are outcomes under the two policy regimes.

- If some parameters are invariant to policy changes, they can be safely transported to different policy environments.
- Structural econometricians search for policy invariant “deep parameters” that can be used to forecast policy changes.



- Under one commonly invoked form of policy invariance, policies keep the potential outcomes unchanged for each person:  
 $Y_0^a = Y_0^b, Y_1^a = Y_1^b$ , but affect costs ( $C^a \neq C^b$ ).
- Such invariance rules out social effects including peer effects and general equilibrium effects.

- Let  $D^a$  and  $D^b$  be the choice taken under each policy regime.
- Invoking invariance of potential outcomes, the observed outcomes under each policy regime are  
 $Y^a = Y_0D^a + Y_1(1 - D^a)$  and  $Y^b = Y_0D^b + (1 - D^b)$ .

- The **Policy Relevant Treatment Effect** (PRTE) is

$$\text{PRTE} = E(Y^a - Y^b).$$

- Comparison of aggregate outcomes under policies “*a*” and “*b*”. PRTE extends ATE by recognizing that policies affect incentives to participate (*C*) but do not force people to participate.
- Only if *C* is very large under *b* and very small under *a*, so there is universal nonparticipation under *b* and universal participation under *a*, would ATE and PRTE be the same parameter.

[Link to Appendix](#)

# Appendix

## Proof

(Keep  $X$  implicit)

$$\begin{aligned}
 E(Y_p) &= \int_0^1 E(Y_p \mid P_p(Z_p) = t) dF_{P_p}(t) \\
 &= \int_0^1 \left[ \int_0^1 [\mathbf{1}_{[0,t]}(u_D) E(Y_{1,p} \mid U_D = u_D) \right. \\
 &\quad \left. + \mathbf{1}_{(t,1]}(u_D) E(Y_{0,p} \mid U_D = u_D)] du_D \right] dF_{P_p}(t) \\
 &= \int_0^1 \left[ \int_0^1 [\mathbf{1}_{[u_D,1]}(t) E(Y_{1,p} \mid U_D = u_D) \right. \\
 &\quad \left. + \mathbf{1}_{(0,u_D)}(t) E(Y_{0,p} \mid U_D = u_D)] dF_{P_p}(t) \right] du_D \\
 &= \int_0^1 [(1 - F_{P_p}(u_D)) E(Y_{1,p} \mid U_D = u_D) \\
 &\quad + F_{P_p|X}(u_D) E(Y_{0,p} \mid U_D = u_D)] du_D.
 \end{aligned}$$

## Proof

- Comparing policy  $p$  to policy  $p'$ ,

$$\begin{aligned} E(Y_p) - E(Y_{p'}) \\ = \int_0^1 \underbrace{E(Y_1 - Y_0 \mid U_D = u_D)}_{\text{MTE}(u_D)} (F_{P_{p'}}(u_D) - F_{P_p}(u_D)) du_D, \end{aligned}$$

which gives the required weights.

- Policies shift the distribution of  $P(Z)$ .
- They keep the distribution of  $Y_1$  and  $Y_0$  unchanged.

## Proof

- This derivation involves changing the order of integration.
- Note that from finiteness of the mean,

$$\begin{aligned} E \left| \mathbf{1}_{[0,t]}(u_D) E(Y_{1,p} \mid U_D = u_D) + \mathbf{1}_{(t,1]}(u_D) E(Y_{0,p} \mid U_D = u_D) \right| \\ \leq E(|Y_1| + |Y_0|) < \infty, \end{aligned}$$

$\therefore$  the change in the order of integration is valid by Fubini's theorem.