# A Running Example Based on: Alternative Methods For Evaluating the Impact of Interventions: An Overview

Excerpt from the Journal of Econometrics, 1985

James J. Heckman & Richard Robb Jr.

Econ 312, Spring 2023



Heckman & Robb

Alternative Methods Extract

For the rest of the course, we will work with a model of program participation for the impact of job training on earnings.

#### 1. Earnings functions

- Assume that individuals experience only one opportunity to participate in training.
- This opportunity occurs in period k.
- Training takes a single period for participants to complete.
- During training, participants earn no labor income.



- Earnings of individual i in period t by  $Y_{it}$ .
- Earnings depend on a vector of observed characteristics, X<sub>it</sub>.
- Post-program earnings (t > k) also depend on a dummy variable, d<sub>i</sub>, which equals one if the *i*th individual participates and is zero if he does not.
- $U_{it}$  represent the error term in the earnings equation and assume that  $E[U_{it}] = 0$ .



• For a linear specification, latent earnings is

$$Y_{it}^* = X_{it}\beta + U_{it},$$

 $\beta$  is a vector of parameters.

• 
$$E(U_{it}|X_{it}) = 0$$
 all  $X_{it}$ .



- β is the coefficient of X in the conditional expectation of Y\* given X.
- Observed earnings  $Y_{it}$ :

$$X_{it}\beta + U_{it} \text{ if } t < k \tag{1}$$
  
$$(1 - d_i)(X_{it}\beta + U_{it}) \text{ if } t = k$$
  
$$X_{it}\beta + U_{it} + \alpha_i d_i \text{ if } t > k$$

- $d_i = 1$  if the person takes training and  $d_i = 0$  otherwise:
- $\alpha$  is one definition of the causal or structural effect of training on earnings.
- Observed earnings are the sum of latent earnings Y<sup>\*</sup><sub>it</sub> and the structural shift term d<sub>i</sub>α that is a consequence of training. Y<sub>it</sub> is thus the sum of two random variables when t > k.

- Problem of selection bias arises because  $d_i$  may be correlated with  $U_{it}$ .
- Consequence of selection decisions by agents. Thus, selection bias is present if

 $E(U_{it}d_i) \neq 0.$ 



## • If $E(U_{it}d_i) \neq 0$ ,

## $E(Y_{it} \mid X_{it}, d_i) \neq X_{it}\beta + d_i\alpha.$



 We will also develop the analysis when α varies among individuals α<sub>i</sub>



### 1.2 Enrollment rules

- The decision to participate in training may be determined by a prospective trainee, by a program administrator or both.
- Whatever the specific content of the rule, it can be described in terms of an index function framework.
- Let *IN<sub>i</sub>* be an index of benefits to the appropriate decision-makers from taking training.
- It is a function of observed (Z<sub>i</sub>) and unobserved by the economist (V<sub>i</sub>) variables.
- Thus

$$IN_i = Z_i \gamma + V_i. \tag{2}$$

- What is observed by the agent is a different matter.
- Z may include X<sub>i</sub>

In terms of this function,

$$d_i = egin{cases} 1 & ext{iff } IN_i > 0 \ 0 & ext{otherwise.} \end{cases}$$

The distribution function of  $V_i$ :  $F(v_i) = \Pr(V_i < v_i)$ .

- V<sub>i</sub>: independently and identically distributed across persons.
- Let  $p = E[d_i] = \Pr[d_i = 1]$  and assume 1 > p > 0.



- If that  $V_i$  is distributed independently of  $Z_i$ ,  $Pr(d_i = 1 | Z_i) = F(-Z_i\gamma)$
- Sometimes called the "propensity score" in statistics (see, e.g., Rosenbaum and Rubin, 1983).



• The condition for the existence of selection bias

$$E(U_{it}d_i) \neq 0$$

may occur because of stochastic dependence between  $U_{it}$  and the unobservable  $V_i$  in equation (2) (selection on the unobservables) or because of stochastic dependence between  $U_{it}$  and  $Z_i$  in equation (2) (selection on observables).



## Simple Behavioral Model



Heckman & Robb

Alternative Methods Extract

- Natural starting point: model of trainee self-selection based on a comparison of the expected value of earnings with and without training.
- For simplicity, assume that training programs accept all applicants (no distinction between ATE and ITT).



- Prospective trainees to discount earnings streams by a common discount factor 1/(1 + r).
- From (1) training raises trainee earnings by  $\alpha$  per period.
- While in training, individual *i* receives a subsidy S<sub>i</sub> which may be negative (so there may be direct costs of program participation).
- Trainees forego income in training period k.
- To simplify the expressions, assume that people live forever.



• As of period k, the present value of earnings for a person who does not receive training is

$$PV_i(0) = E_{k-1}\left(\sum_{j=0}^{\infty} \left(\frac{1}{1+r}\right)^j Y_{i,k+j}\right).$$

- $E_{k-1}$  means that the expectation is taken with respect to information available to the prospective trainee in period k-1.
- The expected present value of earnings for a trainee is

$$PV_i(1) = E_{k-1}\left(S_i + \sum_{j=1}^{\infty} \left(\frac{1}{1+r}\right)^j Y_{i,k+j} + \sum_{j=1}^{\infty} \frac{\alpha}{(1+r)^j}\right)$$



• The risk-neutral wealth-maximizing decision rule is to enroll in the program if  $PV_i(I) > PV_i(0)$  or, letting  $IN_i$  denote the index function in decision rule (2),

$$IN_{i} = PV_{i}(1) - PV_{i}(0) = E_{k-1}[S_{i} - Y_{ik} + \alpha/r], \qquad (3)$$

Decision to train is characterized by

$$d_i = \begin{cases} 1 & \text{iff } E_{k-1}[S_i - Y_{ik} + \alpha/r] > 0 \\ 0 & \text{otherwise.} \end{cases}$$
(4)



Let W<sub>i</sub> be the determinant of the subsidy that the econometrician observes (with associated coefficient φ) and let τ<sub>i</sub> be the part which he does not observe:

$$S_i = W_i \phi + \tau_i.$$

A special case of this model arises when agents possess perfect foresight so that E<sub>k-1</sub>[S<sub>i</sub>] = S<sub>i</sub>, E<sub>k-1</sub>[Y<sub>ik</sub>] = Y<sub>ik</sub> and E<sub>k-1</sub>[α/r] = α/r.



• Collecting terms,

$$d_{i} = \begin{cases} 1 & \text{iff } S_{i} - Y_{ik} + \alpha/r = W_{i}\phi + \alpha/r - X_{ik}\beta + \tau_{i} - U_{ik} > 0\\ 0 & \text{otherwise.} \end{cases}$$
(5)

• Then 
$$(\tau_i - U_{ik}) = V_i$$
 in (2) and  $(W_i, X_{ik})$  corresponds to  $Z_i$  in (2).

Assuming that (W<sub>i</sub>, X<sub>ik</sub>) is distributed independently of V<sub>i</sub> makes (5) a standard discrete choice model.



- Suppose decision rule (5) determines enrollment.
- If the costs of program participation are independent of  $U_{it}$  for all t (so both  $W_i$  and  $\tau_i$  are independent of  $U_{it}$ ), then  $E[U_{it}d_i] = 0$  only if the unobservables in period t are (mean) independent of the unobservables in period k or

$$E[U_{it} \mid U_{ik}] = 0 \text{ for } t > k.$$

• Whether or not  $U_{it}$  and  $d_i$  are uncorrelated hinges on the serial dependence properties of  $U_{it}$ .



• If U<sub>it</sub> is a moving average of order m so

$$U_{it} = \sum_{j=1}^m a_j \varepsilon_{i,t-j},$$

where the  $\varepsilon_{i,t-j}$  are iid, then for t - k > m,  $E[U_{it}d_i] = 0$ .

• On the other hand, if  $U_{it}$  follows a first-order autoregressive scheme, then  $E[U_{it} \mid U_{ik}] \neq 0$  for all finite t and k.



- The enrollment decision rules give context to the selection bias problem.
- The estimators differ greatly in their dependence on particular features of these rules.
- Some estimators do not require that these decision rules be specified at all, while other estimators require a great deal of *a priori* specification of these rules.
- Given the inevitable controversy that surrounds specification of enrollment rules, there is always likely to be a preference by analysts for estimators that require little prior knowledge about the decision rule.
- But this often throws away valuable information and ignores the subjective evaluation implicit in  $d_i = 1$ .

