Notes for Paper on Mis-measured, Binary, Endogenous Regressors

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1 Model and Notation

Fill in material from earlier notes so we have everything in one document!

$$p_{jk}^* = P(T^* = t, Z = k)$$
 $p_{jk} = P(T = t, Z = k)$
 $p_k^* = P(T^* = 1|Z = k)$
 $p_k = P(T = 1|Z = k)$
 $q = P(Z = 1)$

Thus,

$$\begin{split} p_{00}^* &= P(T^* = 0|Z = 0)P(Z = 0) = (1 - p_0^*)(1 - q) \\ &= \left(\frac{1 - p_0 - \alpha_1}{1 - \alpha_0 - \alpha_1}\right)(1 - q) \\ p_{10}^* &= P(T^* = 1|Z = 0)P(Z = 0) = p_0^*(1 - q) \\ &= \left(\frac{p_0 - \alpha_0}{1 - \alpha_0 - \alpha_1}\right)(1 - q) \\ p_{01}^* &= P(T^* = 0|Z = 1)P(Z = 1) = (1 - p_1^*)q \\ &= \left(\frac{1 - p_1 - \alpha_1}{1 - \alpha_0 - \alpha_1}\right)q \\ p_{11}^* &= P(T^* = 1|Z = 1)P(Z = 1) = p_1^*(1 - q) \\ &= \left(\frac{p_1 - \alpha_0}{1 - \alpha_0 - \alpha_1}\right)q \end{split}$$

2 CDF Conditions

2.1 Notation

For $t, Z \in \{0, 1\}$ define

$$F_{tk}^*(\tau) = P(Y \le \tau | T^* = t, Z = k)$$

$$F_{tk}(\tau) = P(Y \le \tau | T = t, Z = k)$$

$$F_k(\tau) = P(Y \le \tau | Z = k)$$

Note that the second two are observed for all t, k while the first is never observed since it depends on the unobserved RV T^* .

2.2 Bounds on α_0, α_1

Case I: No Assumptions on Z We begin by considering the bounds that we can derive for the mis-classification error rates without imposing any conditions on Z. In other words we use only the assumption that the measurement error is non-differential and the structure of the model, namely $Y = \beta T^* + U$.

2.3 Independent Instrument

Assume that $Z \perp U$. The model is $Y = \beta T^* + U$ and

$$F_U(\tau) = P(U \le \tau) = P(Y - \beta T^* \le \tau)$$

but if Z is independent of U then it follows that

$$F_{U}(\tau) = F_{U|Z=k}(\tau) = P(U \le \tau | Z = k) = P(Y - \beta T^* \le \tau | Z = k)$$

$$= P(Y \le \tau | T^* = 0, Z = k)(1 - p_k^*) + P(Y \le \tau + \beta | T^* = 1, Z = k)p_k^*$$

$$= (1 - p_k^*)F_{0k}^*(\tau) + p_k^*F_{1k}^*(\tau + \beta)$$

for all k by the Law of Total Probability. Similarly,

$$F_k(\tau) = (1 - p_k^*) F_{0k}^*(\tau) + p_k^* F_{1k}^*(\tau)$$

and rearranging

$$(1 - p_k^*)F_{0k}^*(\tau) = F_k(\tau) - p_k^*F_{1k}^*(\tau)$$

Substituting this expression into the equation for $F_U(\tau)$ from above, we have

$$F_U(\tau) = F_k(\tau) + p_k^* \left[F_{1k}^*(\tau + \beta) - F_{1k}^*(\tau) \right]$$

for all k and all τ . Evaluating at two values k and ℓ in the support of Z and equating

$$F_k(\tau) + p_k^* \left[F_{1k}^*(\tau + \beta) - F_{1k}^*(\tau) \right] = F_\ell(\tau) + p_\ell^* \left[F_{1\ell}^*(\tau + \beta) - F_{1\ell}^*(\tau) \right]$$

or equivalently

$$F_k(\tau) - F_\ell(\tau) = p_\ell^* \left[F_{1\ell}^*(\tau + \beta) - F_{1\ell}^*(\tau) \right] - p_k^* \left[F_{1k}^*(\tau + \beta) - F_{1k}^*(\tau) \right]$$
 (2.1)

for all τ . Now we simply need to re-express all of the "star" quantities, namely p_k^*, p_ℓ^* and $F_{1k}^*, F_{1\ell}^*$ in terms of α_0, α_1 and the *observable* probability distributions F_{1k} and $F_{1\ell}$

and observable probabilities p_k, p_ℓ . To do this, we use the fact that

$$F_{0k}(\tau) = \frac{1 - \alpha_0}{1 - p_k} (1 - p_k^*) F_{0k}^*(\tau) + \frac{\alpha_1}{1 - p_k} p_k^* F_{1k}^*(\tau)$$

$$F_{1k}(\tau) = \frac{\alpha_0}{p_k} (1 - p_k^*) F_{0k}^*(\tau) + \frac{1 - \alpha_1}{p_k} p_k^* F_{1k}^*(\tau)$$

for all k by Bayes' rule. Solving these equations,

$$p_k^* F_{1k}^*(\tau) = \frac{1 - \alpha_0}{1 - \alpha_0 - \alpha_1} p_k F_{1k}(\tau) - \frac{\alpha_0}{1 - \alpha_0 - \alpha_1} (1 - p_k) F_{0k}(\tau)$$

for all k. Combining this with Equation 2.1, we find that

$$(1 - \alpha_0 - \alpha_1) [F_k(\tau) - F_\ell(\tau)] = \alpha_0 \{ (1 - p_k) [F_{0k}(\tau + \beta) - F_{0k}(\tau)] - (1 - p_\ell) [F_{0\ell}(\tau + \beta) - F_{0\ell}(\tau)] \}$$
$$- (1 - \alpha_0) \{ p_k [F_{1k}(\tau + \beta) - F_{1k}(\tau)] - p_\ell [F_{1\ell}(\tau + \beta) - F_{1\ell}(\tau)] \}$$

Now, define

$$\Delta_{tk}^{\tau}(\beta) = F_{tk}(\tau + \beta) - F_{tk}(\tau) = E\left[\frac{\mathbf{1}\left\{T = t, Z = k\right\}}{p_{tk}} \left(\mathbf{1}\left\{Y \le \tau + \beta\right\} - \mathbf{1}\left\{Y \le \tau\right\}\right)\right]$$

and note that we can express $F_k(\tau) - F_\ell(\tau)$ similarly as

$$F_k(\tau) - F_\ell(\tau) = E\left[\mathbf{1}\left\{Y \le \tau\right\} \left(\frac{\mathbf{1}\left\{Z = k\right\}}{q_k} - \frac{\mathbf{1}\left\{Z = \ell\right\}}{q_\ell}\right)\right]$$

Using this notation, we can write the preceding as

$$(1 - \alpha_0 - \alpha_1) \left[F_k(\tau) - F_\ell(\tau) \right] = \alpha_0 \left[(1 - p_k) \Delta_{0k}^{\tau}(\beta) - (1 - p_\ell) \Delta_{0\ell}^{\tau}(\beta) \right] - (1 - \alpha_0) \left[p_k \Delta_{1k}^{\tau}(\beta) - p_\ell \Delta_{1\ell}^{\tau}(\beta) \right]$$

or in moment-condition form

$$E\left[(1 - \alpha_0 - \alpha_1) \mathbf{1} \left\{ Y \le \tau \right\} \left(\frac{\mathbf{1} \left\{ Z = k \right\}}{q_k} - \frac{\mathbf{1} \left\{ Z = \ell \right\}}{q_\ell} \right) - (\mathbf{1} \left\{ Y \le \tau + \beta \right\} - \mathbf{1} \left\{ Y \le \tau \right\}) \left\{ \alpha_0 \left((1 - p_k) \frac{\mathbf{1} \left\{ T = 0, Z = k \right\}}{p_{0k}} - (1 - p_\ell) \frac{\mathbf{1} \left\{ T = 0, Z = \ell \right\}}{p_{0\ell}} \right) - (1 - \alpha_0) \left(p_k \frac{\mathbf{1} \left\{ T = 1, Z = k \right\}}{p_{1k}} - p_\ell \frac{\mathbf{1} \left\{ T = 1, Z = \ell \right\}}{p_{1\ell}} \right) \right\} \right] = 0$$

Each value of τ yields a moment condition.

3 Special Case: $\alpha_0 = 0$

In this case the expressions from above simplify to

$$(1 - \alpha_1) \left[F_k(\tau) - F_\ell(\tau) \right] + \left\{ p_k \left[F_{1k}(\tau + \beta) - F_{1k}(\tau) \right] - p_\ell \left[F_{1\ell}(\tau + \beta) - F_{1\ell}(\tau) \right] \right\} = 0$$

for all τ .