

1 Data Generating Process

- $T = 2, N = 10000$
- DGP

$$A_{i1} = 1\{\pi_1 Z_{i1} + \alpha_i + v_{i1} \geq 0\}$$

$$Y_{i1} = 1\{\mu_1 A_{i1} + \alpha_i + e_{i1} \geq 0\}$$

$$A_{i2} = 1\{\pi_{21} Y_{i1} + \pi_{22} A_{i1} + \pi_{23} Z_{i2} + \alpha_i + v_{i2} \geq 0\}$$

$$Y_{i2} = 1\{\mu_{21} Y_{i1} + \mu_{22} A_{i2} + \alpha_i + e_{i2} \geq 0\}$$

- $(v_1, e_1, v_2, e_2, \alpha)$, mutually independent, jointly normally distributed, mean γ_1 , variance γ_2
- (Z_1, Z_2) , mutually independent, Bernoulli with β_1 and β_2

2 Linear Programming

2.1 Defining State Variables

- Introduce

$$y_2 = g_2(y_1, a_2, s_2)$$

$$a_2 = h_2(a_1, y_1, z_2, r_2)$$

$$y_1 = g_1(a_1, s_1)$$

$$a_1 = h_1(z_1, r_1)$$

Here, r_1 enumerates the map $z_1 \in \{0, 1\} \mapsto a_1 \in \{0, 1\}$ by enumerating possible value of potential treatment $a_1(z_1)$: $(a_1(0), a_1(1))$. That is,

$$r_1 = \beta(a_1(0), a_1(1))$$

r_1	$a_1(0)$	$a_1(1)$
1	0	0
2	1	0
3	0	1
4	1	1

Table 1: r_1

s_1	$y_1(0)$	$y_1(1)$
1	0	0
2	1	0
3	0	1
4	1	1

Table 2: s_1 as possible states of map $y_1(a_1)$

where $\beta(\cdot) = bi2de(\cdot)$. Likewise, s_1 enumerates the map $a_1 \in \{0, 1\} \mapsto y_1 \in \{0, 1\}$, r_2 enumerates $(a_1, y_1, z_2) \mapsto a_2$, and s_2 enumerates $(y_1, a_2) \mapsto y_2$:

$$\begin{aligned}
s_1 &= \beta(y_1(0), y_1(1)) \\
r_2 &= \beta(a_2(0, 0, 0), a_2(0, 0, 1), \dots, a_2(1, 1, 1)) \\
s_2 &= \beta(y_2(0, 0), y_2(0, 1), \dots, y_2(1, 1))
\end{aligned}$$

Then the corresponding sets have cardinality:

$$\begin{aligned}
|\mathcal{R}_1| &= 2^2 \\
|\mathcal{S}_1| &= 2^2 \\
|\mathcal{R}_2| &= 2^{2^3} = 2^8 \\
|\mathcal{S}_2| &= 2^{2^2} = 2^4
\end{aligned}$$

- Now define

$$q(r_1, s_1, r_2, s_2) \equiv \Pr[r_1, s_1, r_2, s_2]$$

in \mathcal{Q} with $|\mathcal{Q}| = |\mathcal{R}_1| \times |\mathcal{S}_1| \times |\mathcal{R}_2| \times |\mathcal{S}_2| = 2^2 \times 2^2 \times 2^8 \times 2^4 = 65,536$.

- Tables

r_2	$a_2(000)$	$a_2(100)$	$a_2(010)$	$a_2(110)$	$a_2(001)$	$a_2(101)$	$a_2(011)$	$a_2(111)$
1	0	0	0	0	0	0	0	0
2	1	0	0	0	0	0	0	0
3	0	1	0	0	0	0	0	0
4	1	1	0	0	0	0	0	0
5	0	0	1	0	0	0	0	0
6	1	0	1	0	0	0	0	0
7	0	1	1	0	0	0	0	0
8	1	1	1	0	0	0	0	0
\vdots								
256	1	1	1	1	1	1	1	1

Table 3: r_2 as possible states of map $a_2(a_1, y_1, z_2)$

s_2	$y_2(00)$	$y_2(10)$	$y_2(01)$	$y_2(11)$
1	0	0	0	0
2	1	0	0	0
3	0	1	0	0
4	1	1	0	0
5	0	0	1	0
6	1	0	1	0
7	0	1	1	0
8	1	1	1	0
\vdots				
16	1	1	1	1

Table 4: s_2 as possible states of map $y_2(y_1, a_2)$

2.2 Potential Outcomes

- Then, e.g.,

$$\begin{aligned}
& \Pr[Y_1(1) = 1, Y_2(1, 1) = 1] \\
&= \Pr[\mathbf{Y}(1, 1) = (1, 1)] \\
&= \Pr[S_1, S_2 : g_1(1, S_1) = 1, g_2(1, 1, S_2) = 1] \\
&= \sum_{(\beta, \beta') : y_1^1 = y_2^{11} = 1} \Pr[S_1 = \beta(y_1^0, y_1^1), S_2 = \beta'(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})] \\
&= \sum_{(r_1, r_2) \in \mathcal{R}_1 \times \mathcal{R}_2} \sum_{(\beta, \beta') : y_1^1 = y_2^{11} = 1} q[r_1, \beta(y_1^0, y_1^1), r_2, \beta'(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})]
\end{aligned}$$

and

$$\begin{aligned}
& \Pr[Y_1(1) = 0, Y_2(1, 1) = 1] \\
&= \Pr[\mathbf{Y}(1, 1) = (0, 1)] \\
&= \sum_{(\beta, \beta') : y_1^1 = 0, y_2^{01} = 1} \Pr[S_1 = \beta(y_1^0, y_1^1), S_2 = \beta'(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})]
\end{aligned}$$

Likewise,

$$\begin{aligned}
& \Pr[Y_1(0) = 1, Y_2(1, 1) = 1] \\
&= \Pr[\mathbf{Y}(0, 1) = (1, 1)] \\
&= \sum_{(\beta, \beta') : y_1^0 = y_2^{10} = 1} \Pr[S_1 = \beta(y_1^0, y_1^1), S_2 = \beta'(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})]
\end{aligned}$$

and

$$\begin{aligned}
& \Pr[Y_1(1) = 1, Y_2(1, 0) = 1] \\
&= \Pr[\mathbf{Y}(1, 0) = (1, 1)] \\
&= \sum_{(\beta, \beta') : y_1^1 = y_2^{10} = 1} \Pr[S_1 = \beta(y_1^0, y_1^1), S_2 = \beta'(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})]
\end{aligned}$$

And so on...

Regime #	d_1	$d_2(1, d_1)$	$d_2(0, d_1)$
1	0	0	0
2	1	0	0
3	0	1	0
4	1	1	0
5	0	0	1
6	1	0	1
7	0	1	1
8	1	1	1

Table 5: Dynamic Regimes

- Table for regime labels

$$\begin{aligned}
1 : & P[\mathbf{Y}(0, 0) = (1, 1)] + P[\mathbf{Y}(0, 0) = (0, 1)], \\
2 : & P[\mathbf{Y}(1, 0) = (1, 1)] + P[\mathbf{Y}(1, 0) = (0, 1)], \\
3 : & P[\mathbf{Y}(0, 1) = (1, 1)] + P[\mathbf{Y}(0, 0) = (0, 1)], \\
4 : & P[\mathbf{Y}(1, 1) = (1, 1)] + P[\mathbf{Y}(1, 0) = (0, 1)], \\
5 : & P[\mathbf{Y}(0, 0) = (1, 1)] + P[\mathbf{Y}(0, 1) = (0, 1)], \\
6 : & P[\mathbf{Y}(1, 0) = (1, 1)] + P[\mathbf{Y}(1, 1) = (0, 1)], \\
7 : & P[\mathbf{Y}(0, 1) = (1, 1)] + P[\mathbf{Y}(0, 1) = (0, 1)], \\
8 : & P[\mathbf{Y}(1, 1) = (1, 1)] + P[\mathbf{Y}(1, 1) = (0, 1)].
\end{aligned}$$

- When $h_1 > 0$ and $h_2(o_1) > 0 \forall o_1$, then the analytical solution predicts that 7 and 8 are in the ID set

2.3 Constraints

- Number of constraints: $2^6 + 65,536 + 1$
 - but most of them are just defining the simplex
- Constraints: Let

$$p(a_1, y_1, a_2, y_2 | z_1, z_2)$$

be the distribution of the data

- Then, e.g.,

$$\begin{aligned}
p(0, 0, 0, 0|0, 0) &= \Pr[A_1(0) = 0, Y_1(0) = 0, A_2(0, 0, 0) = 0, Y_2(0, 0) = 0|Z_1 = 0, Z_2 = 0] \\
&= \Pr[A_1(0) = 0, Y_1(0) = 0, A_2(0, 0, 0) = 0, Y_2(0, 0) = 0] \\
&= \sum_{(\beta, \beta', \beta'', \beta''') : a_1^0 = y_1^0 = a_2^{000} = y_2^{00} = 0} q[\beta(a_1^0, a_1^1), \beta'(y_1^0, y_1^1), \beta''(a_2^{000}, a_2^{001}, \dots, a_2^{111}), \beta'''(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})]
\end{aligned}$$

and

$$\begin{aligned}
p(1, 1, 1, 1|1, 1) &= \sum_{(\beta, \beta', \beta'', \beta''') : a_1^1 = y_1^1 = a_2^{111} = y_2^{11} = 1} q[\beta(a_1^0, a_1^1), \beta'(y_1^0, y_1^1), \beta''(a_2^{000}, a_2^{001}, \dots, a_2^{111}), \beta'''(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})] \\
&= \Pr[\mathbf{Y}(1, 1) = (1, 1), \mathbf{A}(1, 1) = (1, 1)]
\end{aligned}$$

Note that

$$\begin{aligned}
&\Pr[\mathbf{Y}(1, 1) = (1, 1)] \\
&= \Pr[\mathbf{Y}(1, 1) = (1, 1), A(1, 1) = (0, 0)] + \Pr[\mathbf{Y}(1, 1) = (1, 1), A(1, 1) = (0, 1)] \\
&+ \Pr[\mathbf{Y}(1, 1) = (1, 1), A(1, 1) = (1, 0)] + \Pr[\mathbf{Y}(1, 1) = (1, 1), A(1, 1) = (1, 1)] \\
&= \sum_{\beta : y_1^1 = y_2^{11} = 1} \sum_{\beta : a_1^1 = a_2^{101} = 0} q + \sum_{\beta : y_1^1 = y_2^{11} = 1} \sum_{\beta : a_1^1 = 0, a_2^{101} = 1} q + \sum_{\beta : y_1^1 = y_2^{11} = 1} \sum_{\beta : a_1^1 = 1, a_2^{111} = 0} q + \sum_{\beta : y_1^1 = y_2^{11} = 1} \sum_{\beta : a_1^1 = a_2^{111} = 1} q \\
&= \sum_{\beta : y_1^1 = y_2^{11} = 1} \sum_{(r_1, r_2) \in \mathcal{R}_1 \times \mathcal{R}_2} q[r_1, \beta(y_1^0, y_1^1), r_2, \beta(y_2^{00}, y_2^{01}, y_2^{10}, y_2^{11})]
\end{aligned}$$

and so on...

- Constraints 2 (Simplex):

$$\begin{aligned}
\sum_{r_1, s_1, r_2, s_2} q[r_1, s_1, r_2, s_2] &= 1 \\
q[r_1, s_1, r_2, s_2] &\geq 0 \quad \forall (r_1, s_1, r_2, s_2)
\end{aligned}$$

2.4 More Constraints

- Consider instead

$$y_2 = g_2(y_1, a_2, s_2)$$

$$a_2 = h_2(y_1, z_2, r_2)$$

$$y_1 = g_1(a_1, s_1)$$

$$a_1 = h_1(z_1, r_1)$$

2.5 Population Distribution of Data

- Calculating $p(a_1, y_1, a_2, y_2 | z_1, z_2)$
- Recall

$$A_{i1} = 1\{\pi_1 Z_{i1} + \alpha_i + v_{i1} \geq 0\}$$

$$Y_{i1} = 1\{\mu_1 A_{i1} + \alpha_i + e_{i1} \geq 0\}$$

$$A_{i2} = 1\{\pi_{21} Y_{i1} + \pi_{22} A_{i1} + \pi_{23} Z_{i2} + \alpha_i + v_{i2} \geq 0\}$$

$$Y_{i2} = 1\{\mu_{21} Y_{i1} + \mu_{22} A_{i2} + \alpha_i + e_{i2} \geq 0\}$$

- Then

$$\begin{aligned} p(0, 0, 0, 0 | 0, 0) &= \Pr[A_1(0) = 0, Y_1(0) = 0, A_2(0, 0, 0) = 0, Y_2(0, 0) = 0 | Z_1 = 0, Z_2 = 0] \\ &= \Pr[A_1(0) = 0, Y_1(0) = 0, A_2(0, 0, 0) = 0, Y_2(0, 0) = 0] \\ &= \Pr[\alpha_i + v_{i1} < 0, \alpha_i + e_{i1} < 0, \alpha_i + v_{i2} < 0, \alpha_i + e_{i2} < 0] \end{aligned}$$

and

$$\begin{aligned} p(1, 1, 1, 1 | 1, 1) &= \Pr[A_1(1) = 1, Y_1(1) = 1, A_2(1, 1, 1) = 1, Y_2(1, 1) = 1] \\ &= \Pr[\pi_1 + \alpha_i + v_{i1} \geq 0, \mu_1 + \alpha_i + e_{i1} \geq 0, \pi_{21} + \pi_{22} + \pi_{23} + \alpha_i + v_{i2} \geq 0, \mu_{21} + \mu_{22} + \alpha_i + e_{i2} \geq 0] \end{aligned}$$