

Figure 4: Improved lower bound on the DTE under MTR

where

$$F^{L}(y_{0}, y_{1}) = \max \left\{ 0, F_{0}(a_{0}) + F_{1}(a_{1}) - 1, \theta - \left(F_{0}(a_{0}) - F_{0}(y_{0})\right)^{+} - \left(F_{1}(a_{1}) - F_{1}(y_{1})\right)^{+} \right\},$$

$$F^{L}(y_{0}, y_{1}) = \min \left\{ F_{0}(y_{0}), F_{1}(y_{1}), \theta + \left(F_{0}(y_{0}) - F_{0}(a_{0})\right)^{+} + \left(F_{1}(y_{1}) - F_{1}(a_{1})\right)^{+} \right\}.$$

Suppose that marginal distributions F_0 and F_1 are fixed. Lemma 4 shows that sharp bounds on the joint distribution improve when the values of the joint distribution are known at some fixed points. Note that $P(Y_1 \ge Y_0) = 1$ if and only if $F(y,y) = F_1(y)$ for all $y \in \mathbb{R}$. As illustrated in Figure 3,

$$P(Y_0 > Y_1) = P\left[\bigcup_{y \in \mathbb{R}} \{Y_0 > y, Y_1 < y\}\right].$$

Therefore,

$$P(Y_1 \ge Y_0) = 1$$

$$\iff P(Y_0 > Y_1) = 0$$

$$\iff P(Y_0 > y, Y_1 < y) = 0 \text{ for all } y \in \mathbb{R}$$

$$\iff F(y, y) = F_1(y), \text{ for all } y \in \mathbb{R}.$$

Since for each $y \in \mathbb{R}$ the value of F(y,y) is known from the fixed marginal distribution F_1 under MTR, sharp bounds on the joint distribution can be derived by taking the intersection of the bounds under the restriction $F(y,y) = F_1(y)$ over all $y \in \mathbb{R}$. Technical details are presented in Appendix.