

# Local Average Treatment Effect without Monotonicity

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## INTRODUCTION

- Framework: Local Average Treatment Effect (hereafter LATE)

$$\begin{cases} Y = Y_1 D + Y_0(1 - D) \\ D = D_1 Z + D_0(1 - Z) \end{cases}$$

where  $D, Z$  are binary

- Definition: four strata of people

$D_0$	$D_1$	Type
0	0	Never-takers
0	1	Compliers
1	0	Defiers
1	1	Always-takers

- Motivation

LATE Key Assumption: **Monotonicity**  
(Imbens and Angrist, 1994; Angrist et al., 1996)

$$D_1 \geq D_0 \text{ or } D_0 \geq D_1$$

Under independence, monotonicity and exclusion restriction assumptions,

$$\theta_{IV} = \frac{E[Y|Z=1] - E[Y|Z=0]}{E[D|Z=1] - E[D|Z=0]}$$
 identifies LATE for compliers

- Research Question**

What if monotonicity assumption fails to hold?  
Existence of defiers?

- Existence of defiers (Empirical Evidence)

- Angrist and Evans (1998)

### Key variables

Treatment:  $D=1$  if the household has a third child  
IV:  $Z=1$  if the first two children are of the same sex

- Card (1995)

### Key variables

Treatment:  $D=1$  if if the individual has a college degree  
IV:  $Z=1$  there exists a four-year college in the local labor market where the individual was born

- Related Literature

- LATE framework and relaxations of assumptions  
Imbens and Angrist (1994), Angrist et al. (1996) and Vytlacil (2002)  
Test the validity of IV: Huber et al. (2017), Kitagawa (2015),  
Mourifié and Wan (2017)  
Violation of independence: Kédagni (2021)  
Violation of exclusion restriction: Kédagni and Wu (2023)  
Monotonicity (related): Huber et al. (2017), de Chaisemartin (2017),  
Noack (2021), Dahl, Huber and Mellace (2023)
- Sensitivity analysis in the LATE framework  
Breakdown points: Horowitz and Manski (1995)  
Breakdown frontiers: Masten and Poirier (2020, 2021), Noack (2021)
- Empirical literature: Card (1995), Angrist and Evans (1998)

## MODEL AND RESULTS

### Assumptions

Random Assignment, Relevance, Exclusion Restriction

- For any Borel set  $A \in \mathcal{Y}$

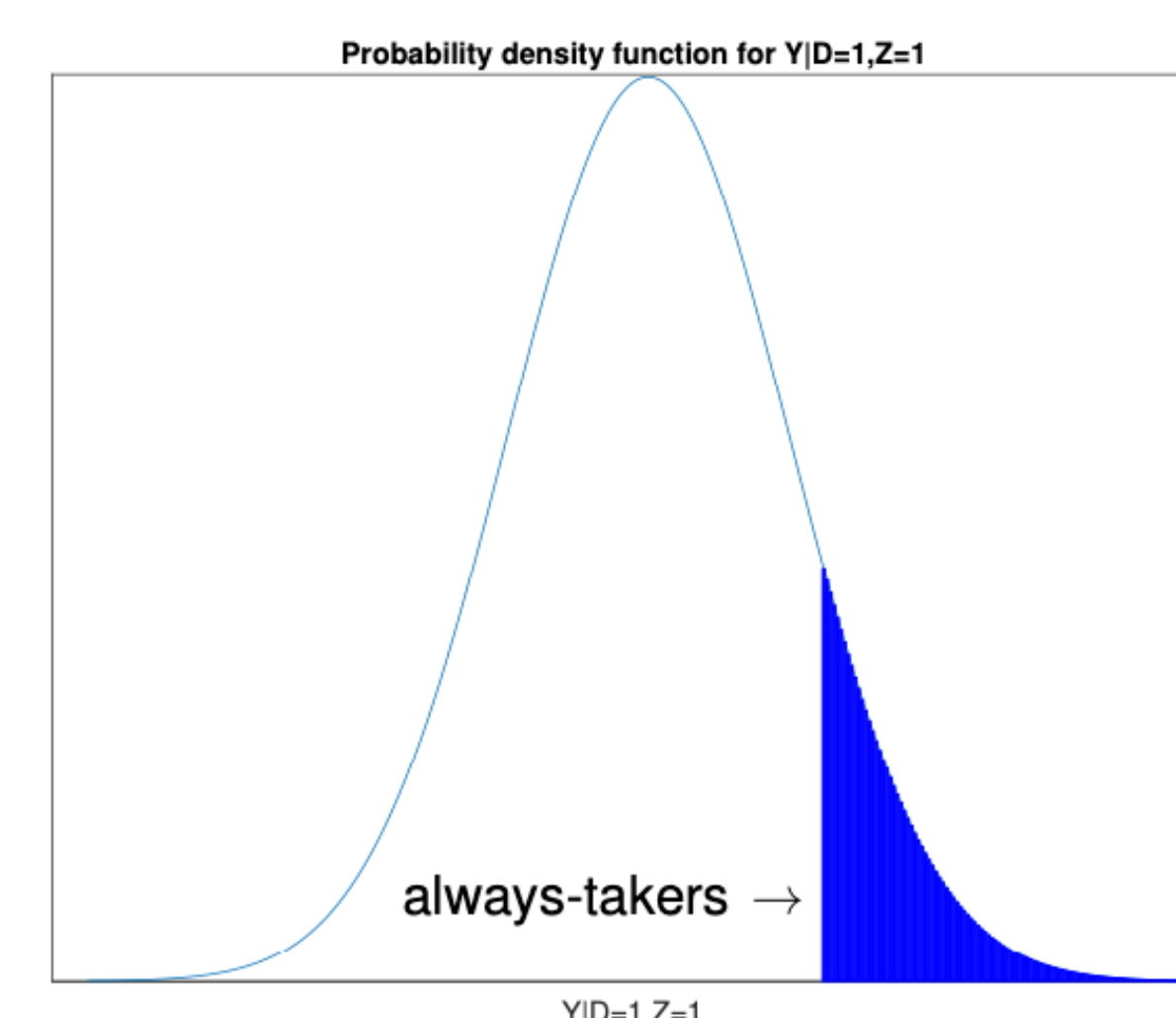
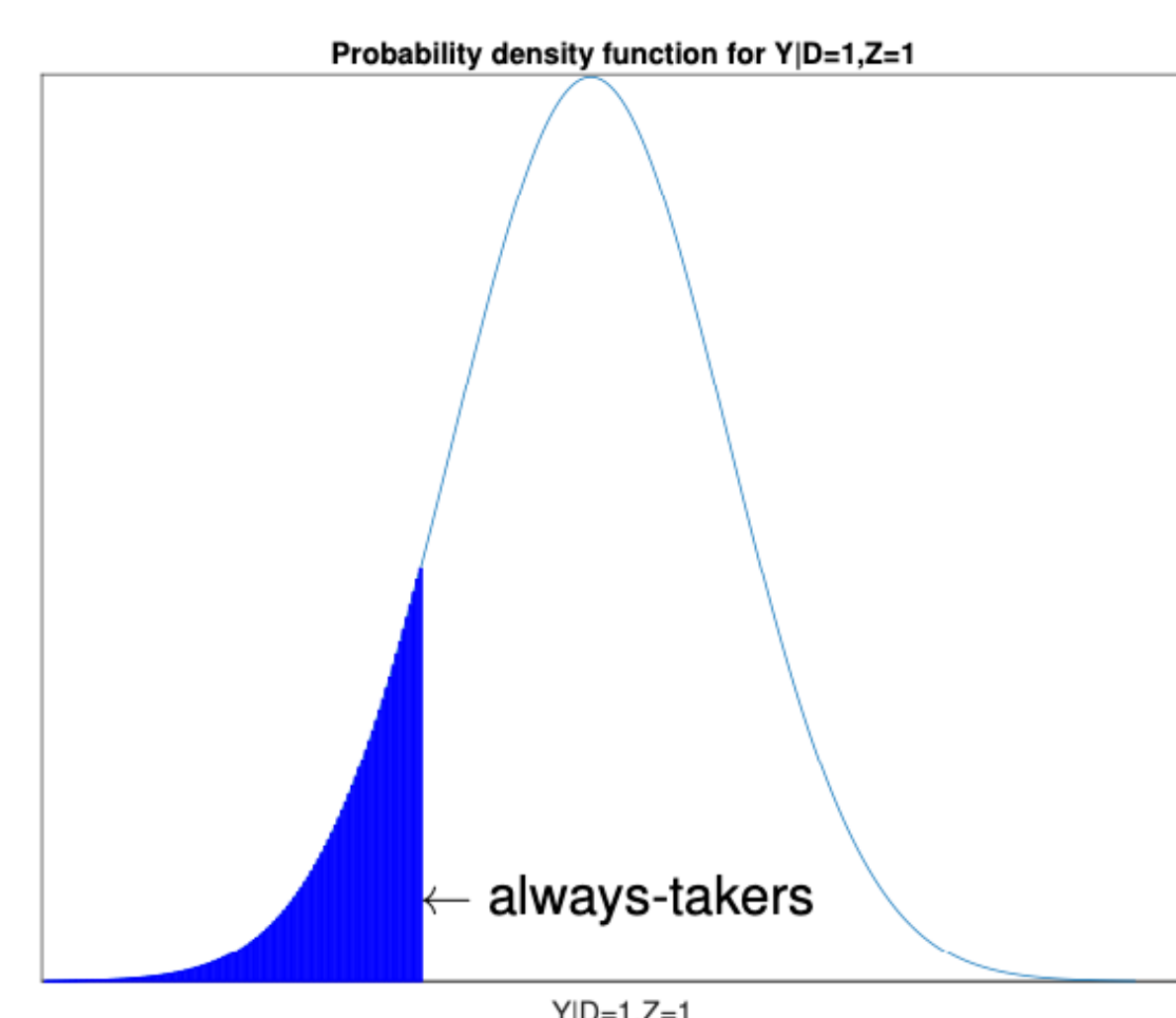
$$\begin{aligned} P(Y \in A, D = 1|Z = 1) &= p_c P(Y_1 \in A|T = c) + p_a P(Y_1 \in A|T = a) \\ P(Y \in A, D = 1|Z = 0) &= p_{df} P(Y_1 \in A|T = df) + p_a P(Y_1 \in A|T = a) \\ P(Y \in A, D = 0|Z = 1) &= p_{df} P(Y_0 \in A|T = df) + p_n P(Y_0 \in A|T = n) \\ P(Y \in A, D = 0|Z = 0) &= p_c P(Y_0 \in A|T = c) + p_n P(Y_0 \in A|T = n) \end{aligned}$$

Proposition 1: Sharp bounds for  $p_{df}$  (Noack, 2021)

$$\max \left\{ \max_{s \in \{0,1\}} \left\{ \sup_A \{P(Y \in A, D = s|Z = 1 - s)\} \right. \right. \\ \left. \left. - (Y \in A, D = s|Z = s) \right\}, 0 \right\} \\ \leq p_{df} \leq \min \{E[D|Z = 0], E[1 - d|Z = 1]\}$$

Lemma 1: Mixture Representation

$$\begin{aligned} E[Y|D = 1, Z = 1] &= \frac{p_a}{E[D|Z = 1]} \mu_{1a} + \frac{E[D|Z = 1] - p_a}{E[D|Z = 1]} \mu_{1c} \\ E[Y|D = 1, Z = 0] &= \underbrace{\frac{p_a}{E[D|Z = 0]}}_{p_1} \mu_{1a} + \underbrace{\frac{E[D|Z = 0] - p_a}{E[D|Z = 0]}}_{1-p_1} \mu_{1df} \end{aligned}$$



Proposition 2: Sharp bounds for  $\mu_{1a}(p_a)$

$$\max\{LB_{1a}^1(p_a), LB_{1a}^0(p_a)\} \leq \mu_{1a}(p_a) \leq \min\{UB_{1a}^1(p_a), UB_{1a}^0(p_a)\}$$

Theorem 1: Sharp bounds for  $\mu_{1c}$  and  $\mu_{1df}$

$$\begin{aligned} \mu_{1c} &\in \left[ \inf_{p_a \in \mathcal{Y}, \mu_{1a} \in \Gamma_1(p_a)} \left\{ \frac{E[YD|Z = 1] - p_a \mu_{1a}}{E[D|Z = 1] - p_a} \right\}, \sup_{p_a \in \mathcal{Y}, \mu_{1a} \in \Gamma_1(p_a)} \left\{ \frac{E[YD|Z = 1] - p_a \mu_{1a}}{E[D|Z = 1] - p_a} \right\} \right] \\ \mu_{1df} &\in \left[ \inf_{p_a \in \mathcal{Y}, \mu_{1a} \in \Gamma_1(p_a)} \left\{ \frac{E[YD|Z = 0] - p_a \mu_{1a}}{E[D|Z = 0] - p_a} \right\}, \sup_{p_a \in \mathcal{Y}, \mu_{1a} \in \Gamma_1(p_a)} \left\{ \frac{E[YD|Z = 0] - p_a \mu_{1a}}{E[D|Z = 0] - p_a} \right\} \right] \end{aligned}$$

## ILLUSTRATIONS

- DGP 1

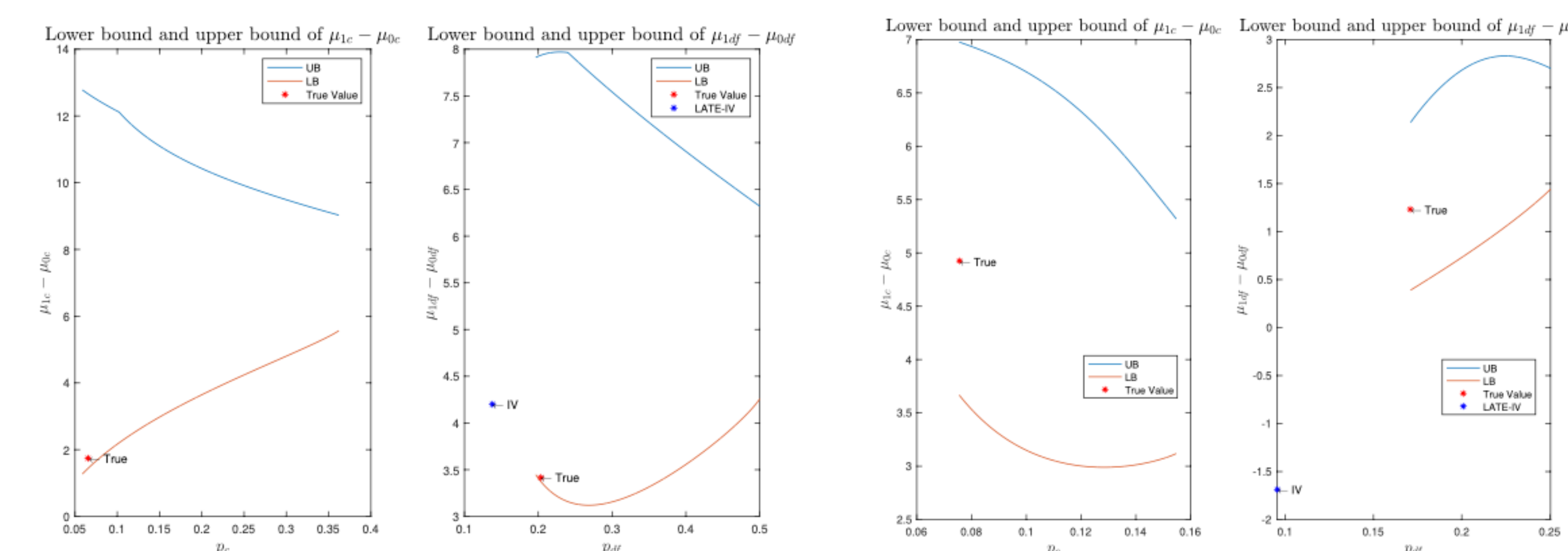
$$\begin{cases} Y = \beta D + U \\ D = 1\{V > \delta Z\} \\ Z = 1\{\varepsilon > 0\} \end{cases}$$

where  $\beta = 5\Phi(\theta)$ ,  $U = 2\theta$ ,  $V = \theta$ ,  $(\theta, \delta, \varepsilon)' \sim N(0, [0, 0.5, 0]')$ ,  $\Phi(\cdot)$  is the standard normal CDF.

- DGP 2

$$\begin{cases} Y = \beta D + U \\ D = 1\{V_1 > 2Z, V_2 > Z\}, \\ Z = 1\{\varepsilon > 0\} \end{cases}$$

where  $\beta = 5\Phi(2V_1 + V_1)$ ,  $U = \frac{1}{2}(V_1 + V_2)$ ,  $(V_1, V_2, \varepsilon)' \sim N(0, I)$ ,  $\Phi(\cdot)$  is the standard normal CDF.

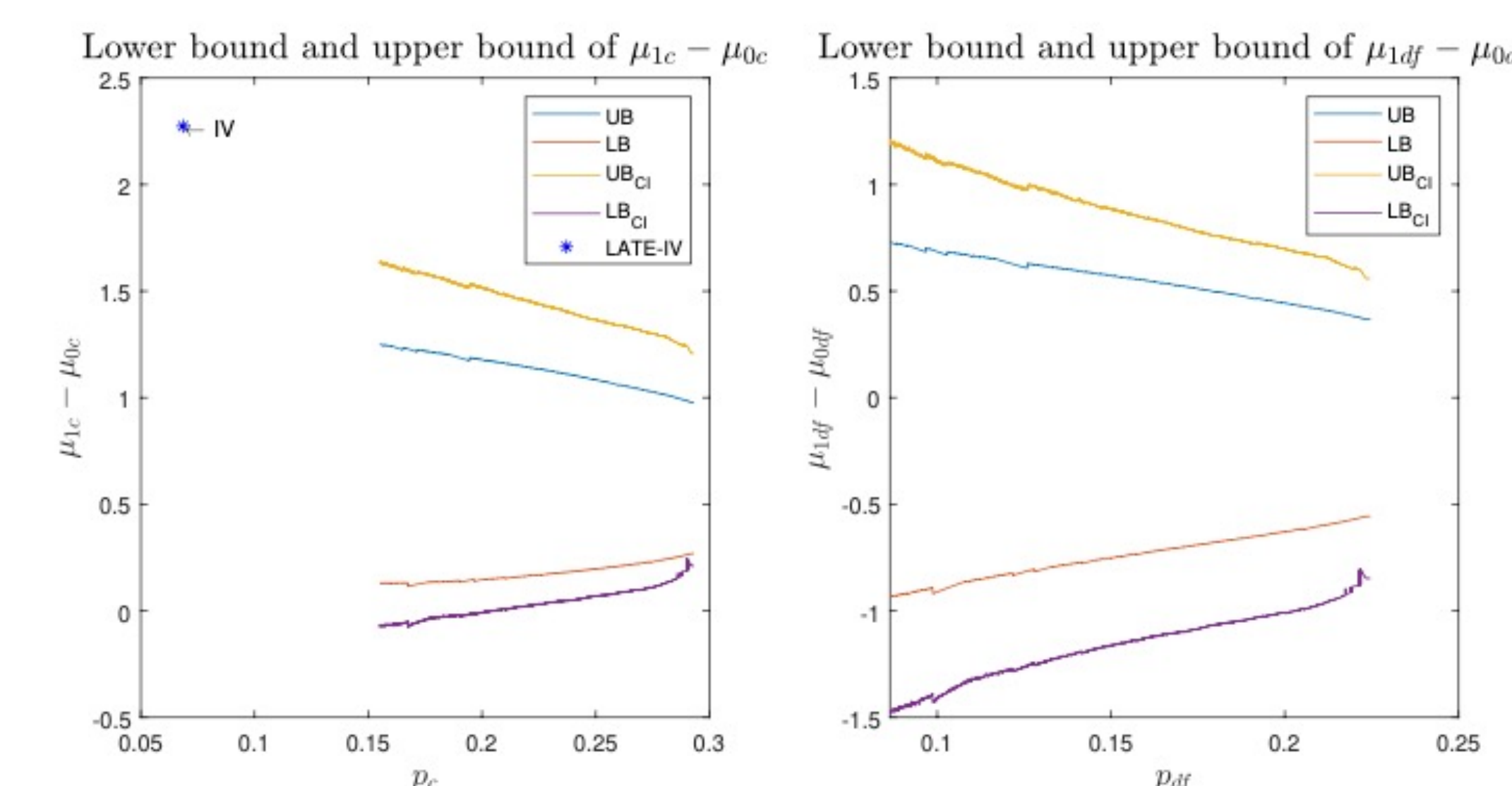


DGP 1: Bounds for LATE with IV and true value

DGP 2: Bounds for LATE with IV and true value

- Empirical examples

- Card (1995)



- Angrist and Evans (1998)

