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Introduction

AER publication (2005-2025)

Study	Mono	Indep	IV	FRD
Allcott et al. (2020)		✓	✓	
Andrews (2016)	✓	✓		✓
Autor et al. (2019)	✓	✓	✓	
Dinkelman (2011)		✓	✓	
...
Simcoe (2012)			✓	

Table: Summary statistics of AER using IV/FRD (2005-2025)¹

1. Note: Only 19 of these 39 articles (50%) that explicitly identify a local average treatment effect using either an IV or FRD approach include the word
“monotonicity”!

Definition

Framework (FRD and LATE)

Let (Ω, \mathcal{F}, P) be the probability space

$D(\cdot, \cdot) : \mathcal{R} \times \Omega \rightarrow \{0, 1\}$ is the observed binary treatment assignment,

$Y(\cdot, \cdot) : \mathcal{R} \times \Omega \rightarrow \mathcal{Y}$ is the observed outcome of interest,

$R(\cdot) : \Omega \rightarrow \mathcal{R}$ is a continuous running variable with a known cut-off r_0 .

For the outcome $Y(R)$, we have $Y(R) := Y_1(R)D(R) + Y_0(R)(1 - D(R))$.

(D is binary.)

Motivation

Key Assumption: Local Monotonicity (Arai et al., 2022 and Hsu et al., 2024)

There exists a small $\epsilon > 0$ such that $T_\epsilon \in \{a, c, n\}$ almost surely.

- Also called no-defier assumption
- Under local continuity and local monotonicity,
 - $\theta_{FRD} = \frac{\mathbb{E}[Y|R=r_0^+] - \mathbb{E}[Y|R=r_0^-]}{\mathbb{E}[D|R=r_0^+] - \mathbb{E}[D|R=r_0^-]}$ identifies LATE for compliers (Hahn et al, 2001), where R is the running variable and r_0 is the cutoff point.
- **Question**
 - Existence of defiers?
 - What if local monotonicity fails to hold?

Existence of defiers (Kirkeboen et al, 2016)

Key variables (Kirkeboen et al, 2016)

Treatment : $D=1$ if student graduates with a degree in science rather than a degree in humanities

R : student's performance score for college admission, the cut-off is an admission threshold to a competitive major (say, science) rather than less competitive majors (say, humanities)

- **Existence of defiers** (Arai et al., 2022)
 - if some students, who tend to be attracted by nonmajored subjects and/or change their minds about their career choices, always switch from their assigned major to the other based on revisions of their beliefs or preferences

Literature Review

- LATE framework literature and relaxations of assumptions
 - Imbens and Angrist (1994), Angrist et al. (1996) and Vytlacil (2002)
 - Test the validity of IV: Huber and Mellace (2015), Kitagawa (2015), Mourifié and Wan (2017)
 - Independence: Kédagni (2023)
 - Exclusion restriction: [Cui et al. \(2024\)](#)
 - Monotonicity (related): Huber et al. (2017), de Chaisemartin (2017), Noack (2021), Fiorini and Stevens (2021), [Cui et al. \(2024\)](#)
- RD framework literature (SRD and FRD)
 - Thistlethwaite and Campbell (1960): SRD
 - RD Application: Angrist and Lavy (1999), van der Klaauw (2002), Lee et al. (2004), DiNardo and Lee (2011), Choi and Lee (2023)
 - FRD framework: Arai et al. (2022), Hsu et al. (2024)
 - Local continuity and local monotonicity

Model and Identification

Definition (Four strata of people add “i”)

$$T_{\epsilon} = \begin{cases} a, & \text{if } D(r) = 1, \text{ for } r \in B_{\epsilon}, \\ n, & \text{if } D(r) = 0, \text{ for } r \in B_{\epsilon}, \\ c, & \text{if } D(r) = 1 \{r \geq r_0\}, \text{ for } r \in B_{\epsilon}, \\ df, & \text{if } D(r) = 1 \{r < r_0\}, \text{ for } r \in B_{\epsilon}, \\ i, & \text{otherwise, } \quad \text{► Existence Example} \end{cases} \quad (1)$$

where $B_{\epsilon} = \{r \in \mathcal{R} : |r - r_0| \leq \epsilon\}$ of the cutoffs.

Model Setting

Assumption 1. Local continuity in distributions

For $d \in \{0, 1\}$, $t \in \{a, c, n, df\}$, and all measurable subset $A \subseteq \mathcal{Y}$,

$$\lim_{\varepsilon \rightarrow 0} \mathbb{P}(Y_d \in A, T_\varepsilon = t \mid R = r_0 - \varepsilon) = \lim_{\varepsilon \rightarrow 0} \mathbb{P}(Y_d \in A, T_\varepsilon = t \mid R = r_0 + \varepsilon). \quad (2)$$

Assumption 2: Local no indefinite type

For $t = i$,

$$\lim_{\varepsilon \rightarrow 0} P(T_\varepsilon = t \mid Z = c - \varepsilon) = \lim_{\varepsilon \rightarrow 0} P(T_\varepsilon = t \mid Z = c + \varepsilon) = 0 \quad (3)$$

Model Setting

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

$$\begin{aligned}\mathbb{P}(Y \in A, D = 1 | R = r_0^+) &= \mathbb{P}(T = c | R = r_0^+) \mathbb{P}(Y_1(r_0) \in A | T = c, R = r_0^+) \\ &\quad + \mathbb{P}(T = a | R = r_0^+) \mathbb{P}(Y_1(r_0) \in A | T = a, R = r_0^+)\end{aligned}$$

$$\begin{aligned}\mathbb{P}(Y \in A, D = 1 | R = r_0^-) &= \mathbb{P}(T = df | R = r_0^-) \mathbb{P}(Y_1(r_0) \in A | T = df, R = r_0^-) \\ &\quad + \mathbb{P}(T = a | R = r_0^-) \mathbb{P}(Y_1(r_0) \in A | T = a, R = r_0^-)\end{aligned}$$

$$\begin{aligned}\mathbb{P}(Y \in A, D = 1 | R = r_0) &= \mathbb{P}(T = a | R = r_0) \cdot \mathbb{P}(Y_1(r_0) \in A | T = a, R = r_0) \\ &\quad + \mathbb{P}(T = c | R = r_0) \cdot \mathbb{P}(Y_1(r_0) \in A | T = c, R = r_0)\end{aligned}$$

Similar results hold for $D = 0$. [► Visualization](#)

LATE

Mean Potential Outcomes: $\mathbb{E}[Y_1(r)|T = c, R = r]$

$$\begin{aligned}\mathbb{E}[Y_1(r)|T = c, R = r] = & \lambda_c \times \lim_{r \rightarrow r_0} \mathbb{E}[Y_1(r)|T_{0 < |r - r_0| \leq \epsilon} = c, R \in B_\epsilon \setminus r_0] \\ & + (1 - \lambda_c) \times \mathbb{E}[Y_1(r)|T_{r=r_0} = c, R = r_0],\end{aligned}$$

where

$$\lambda_c = \frac{\mathbb{P}(R \in B_\epsilon \setminus \{r_0\}, T = c)}{\mathbb{P}(R \in B_\epsilon \setminus \{r_0\}, T = c) + \mathbb{P}(R = r_0, T = c, D = 1)}$$

and

$$\underline{\lambda_c} = \frac{\mathbb{P}(R \in B_\epsilon \setminus \{r_0\}, T = c)}{\mathbb{P}(R \in B_\epsilon \setminus \{r_0\}, T = c) + \mathbb{P}(R = r_0, D = 1)}.$$

Conclusion

Main contributions and strengths:

- 1 Extend the LATE framework without monotonicity to FRD
- 2 Derive the sharp bounds for LATE for compliers and defiers

Future work:

- 1 Finish the empirical examples
- 2 Finish the inference part (Chernozhukov et al. (2011) precision-corrected estimator)

Thanks!

Appendix

Definition (cont')

For the existence of type “indefinite”, we can consider the single-threshold crossing specification of potential treatment like [Hsu et al. \(2024\)](#),

$$D(r) = 1\{r + 1\{r \geq r_0\} + X < 0\}, r \in [r_0 - \epsilon, r_0 + \epsilon]$$

where $X \sim N(0, 1)$. For simplicity, we fix $\epsilon > 0$ and $r_0 = 0$, the support of X can be divided into the groups from figure below. [▶ back](#)



Figure: The distribution of types under the support of X

Visualization of Types

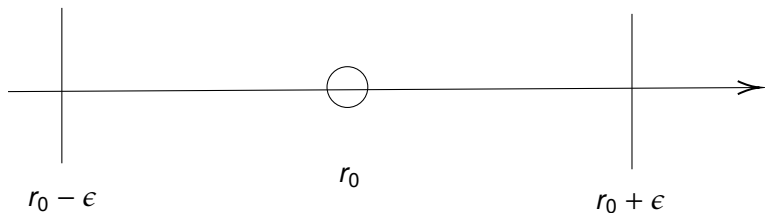
 $D = 1$ a/df a/c a/c $D = 0$ n/c n/df n/df 

Figure: Visualization of types

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