

# **ECMA31100 Introduction to Empirical Analysis II**

Winter 2022, Week 3: Discussion Session

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January 26, 2022

# Welcome to week 3 of ECMA31100!

## Problem set

- Please remember to submit your problem set on time!
- Check the latest announcement post on Canvas

## TA sessions: new time and location

- **Time:** Wednesdays, 3:30pm – 4:20pm
- **Location:** HM 130

## My office hours

- **Time:** Wednesdays, 5:30pm – 7:30pm
- **Location:** Zoom, see link on Canvas
- Feel free to send me an e-mail if you plan to come, so I can prepare ahead

# Review and plans

## Last TA session

- Some general advice and resources for further readings
- Imputation methods
- Matching estimators (matching on covariates and propensity scores)

## Plans for today

- Finish our discussion on blocking (see my slides from week 2)
- Discuss [Abadie \(2003\)](#)

Please let me know if you have any suggestions or if you find any typos in the slides

Any questions?

# Contents

1. Abadie (2003)'s  $\kappa$

2. Proofs

# Model and terminology

## Notations

- Outcome:  $Y$
- Binary treatment:  $D \in \{0, 1\}$
- Binary instrument:  $Z \in \{0, 1\}$

## Potential outcomes

- $Y = DY_1 + (1 - D)Y_0$ , where  $Y_d$  is the potential outcome if  $D = d$
- $D = ZD_1 + (1 - Z)D_0$ , where  $D_z$  is the potential outcome if  $Z = z$

## Four groups

Always-takers (a)	Compliers (c)	Never-takers (n)	Defiers (d)
$D_1 = D_0 = 1$	$D_1 = 1$ and $D_0 = 0$	$D_1 = D_0 = 0$	$D_1 = 0$ and $D_0 = 1$

# Abadie (2003)'s $\kappa$

## Assumptions

- Independence:  $(Y_1, Y_0, D_1, D_0) \perp\!\!\!\perp Z|X$
- Exclusion:  $\mathbb{P}[Y_{D_1} = Y_{D_0}|X] = 1$
- First stage:  $0 < \mathbb{P}[Z = 1|X] < 1$  and  $\mathbb{P}[D_1 = 1|X] > \mathbb{P}[D_0 = 1|X]$
- Monotonicity:  $\mathbb{P}[D_1 \geq D_0|X] = 1$

## Theorem 3.1 of Abadie (2003)

- For any measurable  $f(Y, X, D)$  such that  $\mathbb{E}|f(Y, D, X)| < \infty$ , we have

$$\mathbb{E}[f(Y, X, D)|D_1 > D_0] = \frac{1}{\mathbb{P}[D_1 > D_0]} \mathbb{E}[\kappa f(Y, X, D)],$$

where

$$\kappa \equiv 1 - \frac{D(1-Z)}{\mathbb{P}[Z=0|X]} - \frac{(1-D)Z}{\mathbb{P}[Z=1|X]}$$

# Abadie (2003)'s $\kappa$

## Some intuition

- We cannot observe both  $D_0$  and  $D_1 \implies$  Compliers are not identified individually
- Distribution of observables can be described for the population of the compliers
- $\kappa$ : Compliers = 1 – Always takers – Never takers

## Proof strategy

► Proof

- **Notations:** Let  $G \in \{a, c, n\}$  (defiers are ruled out by monotonicity)
  - $G = a$  corresponds to  $D_0 = D_1 = 1$
  - $G = c$  corresponds to  $D_1 = 1 > 0 = D_0$
  - $G = n$  corresponds to  $D_0 = D_1 = 0$
- **Step 1:** Show that  $\mathbb{E}[\kappa | G, X, D, Y] = \mathbb{1}[G = c]$
- **Step 2:** Use the law of iterated expectations on  $\mathbb{E}[\kappa f(Y, X, D)]$



Any questions?

# Abadie (2003)'s $\kappa$

## Estimation: least squares

### General set-up

- Let  $\theta_0$  be the vector of parameters such that  $\mathbb{E}[Y|X, D, D_1 > D_0] = h(D, X, \theta_0)$
- Then,  $\theta_0 = \arg \min_{\theta \in \Theta} \mathbb{E}[(Y - h(D, X, \theta))^2 | D_1 > D_0]$  but we cannot identify the compliers
- By the last slide, the above is equivalent to  $\theta_0 = \arg \min_{\theta \in \Theta} \mathbb{E}[\kappa(Y - h(D, X, \theta))^2]$
- The sample analog is  $\hat{\theta} = \arg \min_{\theta \in \Theta} \frac{1}{n} \sum_{i=1}^n \kappa_i [y_i - h(d_i, x_i, \theta)]^2$

### Estimation of $\kappa_i$ (not just for least squares)

- Each  $\kappa_i$  is estimated by  $\kappa_i = 1 - \frac{d_i(1-z_i)}{\hat{\mathbb{P}}[Z=0|x_i]} - \frac{(1-d_i)z_i}{\hat{\mathbb{P}}[Z=1|x_i]}$
- We can use probit to estimate  $\hat{\mathbb{P}}[Z = 0|X]$  and  $\hat{\mathbb{P}}[Z = 1|X]$

# Abadie (2003)'s $\kappa$

Estimation: parameterization for least squares

## Least squares: linear parametrization

- Suppose that  $h(D, X, \theta) = \alpha D + X' \beta$  where  $\theta = (\alpha, \beta)$
- Then, we need to solve:  $(\hat{\alpha}, \hat{\beta}) = \arg \min_{(\alpha, \beta) \in \Theta} \frac{1}{n} \sum_{i=1}^n \kappa_i (y_i - \alpha d_i - x_i' \beta)^2$

## Least squares: probit

- Why probit? Make sure the  $\Phi(\cdot) \in [0, 1]$
- Suppose that  $h(D, X, \theta) = \Phi(\alpha D + X' \beta)$  where  $\theta = (\alpha, \beta)$  and  $\Phi(\cdot)$  is the normal cdf
- Then, we need to solve:  $(\hat{\alpha}, \hat{\beta}) = \arg \min_{(\alpha, \beta) \in \Theta} \frac{1}{n} \sum_{i=1}^n \kappa_i [y_i - \Phi(\alpha d_i + x_i' \beta)]^2$

# Abadie (2003)'s $\kappa$

Estimation: maximum likelihood

## General set-up

- Let  $f(Y, D, X, \theta)$  be the density, then we solve  $\theta_0 = \arg \max_{\theta \in \Theta} \mathbb{E}[\ln f(Y, D, X, \theta) | D_1 > D_0]$
- Again, the above is equivalent to  $\theta_0 = \arg \max_{\theta \in \Theta} \mathbb{E}[\kappa \ln f(Y, D, X, \theta)]$
- The sample analog is  $\hat{\theta} = \arg \max_{\theta \in \Theta} \frac{1}{n} \sum_{i=1}^n \kappa_i \ln f(y_i, d_i, x_i, \theta)^2$

## One possible parameterization

- We may specify  $f$  as  $f(Y, D, X, \beta) = \Phi(\alpha D + X' \beta)^Y [1 - \Phi(\alpha D + X' \beta)]^{1-Y}$  as  $Y$  is binary
- Then, solve  $(\hat{\alpha}, \hat{\beta}) = \arg \max_{(\alpha, \beta) \in \Theta} \frac{1}{n} \sum_{i=1}^n \kappa_i \{y_i \ln \Phi(\alpha d_i + x_i' \beta) + (1 - y_i) \ln [1 - \Phi(\alpha d_i + x_i' \beta)]\}$

# Abadie (2003)'s $\kappa$

## Empirical application

### Question: Impact of 401(k) retirement programs on savings

- Only workers in firms that offer 401(k) are eligible
- Do these tax-deferred retirement plans imply:
  - More savings?
  - Crowd out other types of savings?

### Some discussion on the assumptions

- 401(k) eligibility is ignorable given some observables  $\implies$  Can be used as IV
  - Poterba et al. (1994, 1995): eligibility could be unrelated to saving preferences
- Only eligible individuals can open a 401(k) account  $\implies$  Monotonicity holds

# Abadie (2003)'s $\kappa$

## Empirical application

**Table 1:** Summary statistics (taken from table 1 of Abadie (2000))

	Entire Sample	By 401(k) participation		By 401(k) eligibility	
		Participants	Non-participants	Eligibles	Non-eligibles
<i>Treatment:</i>					
Participation in 401(k)	0.28 (0.45)			0.70 (0.46)	0.00 (0.00)
<i>Instrument:</i>					
Eligibility for 401(k)	0.39 (0.49)	1.00 (0.00)	0.16 (0.37)		
<i>Outcome variables:</i>					
Family Net Financial Assets	19,071.68 (63,963.84)	38,472.96 (79,271.08)	11,667.22 (55,289.23)	30,535.09 (75,018.98)	11,676.77 (54,420.17)
Participation in IRA	0.25 (0.44)	0.36 (0.48)	0.21 (0.41)	0.32 (0.47)	0.21 (0.41)
<i>Covariates:</i>					
Family Income	39,254.64 (24,090.00)	49,815.14 (26,814.24)	35,224.25 (21,649.17)	47,297.81 (25,620.00)	34,066.10 (21,510.64)
Age	41.08 (10.30)	41.51 (9.65)	40.91 (10.53)	41.48 (9.61)	40.82 (10.72)
Married	0.63 (0.48)	0.70 (0.46)	0.60 (0.49)	0.68 (0.47)	0.60 (0.49)
Family Size	2.89 (1.53)	2.92 (1.47)	2.87 (1.55)	2.91 (1.48)	2.87 (1.56)

# Abadie (2003)'s $\kappa$

## Empirical application

**Table 2:** Summary statistics (taken from table 2 of Abadie (2000))

Dependent Variable: Family Net Financial Assets (in \$)				
	Ordinary Least Squares (1)	Endogenous Treatment		
		Two Stage Least Squares		Least Squares Treated (4)
		First Stage (2)	Second Stage (3)	
Participation in 401(k)	13,527.05 (1,810.27)		9,418.83 (2,152.89)	10,800.25 (2,261.55)
Constant	-23,549.00 (2,178.08)	-0.0306 (0.0087)	-23,298.74 (2,167.39)	-27,133.56 (3,212.35)
Family Income (in thousand \$)	976.93 (83.37)	0.0013 (0.0001)	997.19 (83.86)	982.37 (106.65)
Age (minus 25)	-376.17 (236.98)	-0.0022 (0.0010)	-345.95 (238.10)	312.30 (371.76)
Age (minus 25) square	38.70 (7.67)	0.0001 (0.0000)	37.85 (7.70)	24.44 (11.40)
Married	-8,369.47 (1,829.93)	-0.0005 (0.0079)	-8,355.87 (1,829.67)	-6,646.69 (2,742.77)
Family Size	-785.65 (410.78)	0.0001 (0.0024)	-818.96 (410.54)	-1,234.25 (647.42)
Eligibility for 401(k)		0.6883 (0.0080)		

# Abadie (2003)'s $\kappa$

## Empirical application

**Table 3:** Summary statistics (taken from table 3 of Abadie (2000))

Dependent Variable: IRA Account

	Linear Response			Probit Response				
	Least Sq.	Endogenous Treatment		Probit	Least Sq.	Endogenous Treatment		
		Two Stage Least Sq.	Least Sq. Treated			Bivariate Probit	Probit Treated	Least Sq. Treated
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Participation in 401(k)	0.0569 (0.0103)	0.0274 (0.0132)	0.0253 (0.0131)	0.0712 (0.0121)	0.0699 (0.0126)	0.0407 (0.0156)	0.0358 (0.0161)	0.0264 (0.0172)
Family Income (in thousand \$)	0.0059 (0.0002)	0.0060 (0.0002)	0.0060 (0.0003)	0.0069 (0.0003)	0.0070 (0.0003)	0.0069 (0.0003)	0.0069 (0.0004)	0.0072 (0.0005)
Age (minus 25)	0.0074 (0.0014)	0.0076 (0.0014)	0.0119 (0.0025)	0.0149 (0.0022)	0.0153 (0.0023)	0.0147 (0.0021)	0.0183 (0.0034)	0.0207 (0.0037)
Age (minus 25) square	0.0000 (0.0000)	0.0000 (0.0000)	-0.0001 (0.0001)	-0.0001 (0.0001)	-0.0001 (0.0001)	-0.0001 (0.0001)	-0.0002 (0.0001)	-0.0002 (0.0001)
Married	0.0312 (0.0110)	0.0313 (0.0110)	0.0440 (0.0184)	0.0590 (0.0152)	0.0477 (0.0166)	0.0577 (0.0148)	0.0627 (0.0231)	0.0535 (0.0244)
Family Size	-0.0264 (0.0032)	-0.0266 (0.0032)	-0.0340 (0.0053)	-0.0424 (0.0050)	-0.0403 (0.0056)	-0.0415 (0.0049)	-0.0472 (0.0075)	-0.0480 (0.0082)



Any questions?

# Contents

1. Abadie (2003)'s  $\kappa$

2. Proofs

## Proof

- Consider the case with  $G = c$ :

$$\begin{aligned}\mathbb{E}[\kappa | G = c, X, D, Y] &= \mathbb{E} \left[ 1 - \frac{D(1-Z)}{\mathbb{P}[Z=0|X]} - \frac{(1-D)Z}{\mathbb{P}[Z=1|X]} \middle| G = c, X, D, Y \right] \\ &= \mathbb{E} [1 - 0 - 0 | G = c, X, D, Y] \\ &= 1\end{aligned}$$

## Proof (Cont'd)

- Consider the case with  $G = a$ :

$$\begin{aligned}\mathbb{E}[\kappa | G = a, X, D, Y] &= \mathbb{E} \left[ 1 - \frac{D(1-Z)}{\mathbb{P}[Z=0|X]} - \frac{(1-D)Z}{\mathbb{P}[Z=1|X]} \middle| G = a, X, D, Y \right] \\ &= \mathbb{E} \left[ 1 - \frac{1-Z}{\mathbb{P}[Z=0|X]} - 0 \middle| G = a, X, D, Y \right] \\ &= 1 - \mathbb{E} \left[ \frac{1-Z}{\mathbb{P}[Z=0|X]} \middle| X \right] \\ &= 1 - \frac{\mathbb{P}[Z=0|X]}{\mathbb{P}[Z=0|X]} \\ &= 0\end{aligned}$$

## Proof (Cont'd)

- Consider the case with  $G = n$ :

$$\begin{aligned}\mathbb{E}[\kappa | G = n, X, D, Y] &= \mathbb{E} \left[ 1 - \frac{D(1-Z)}{\mathbb{P}[Z=0|X]} - \frac{(1-D)Z}{\mathbb{P}[Z=1|X]} \middle| G = n, X, D, Y \right] \\ &= \mathbb{E} \left[ 1 - 0 - \frac{Z}{\mathbb{P}[Z=1|X]} \middle| G = n, X, D, Y \right] \\ &= 1 - \mathbb{E} \left[ \frac{Z}{\mathbb{P}[Z=1|X]} \middle| X \right] \\ &= 1 - \frac{\mathbb{P}[Z=1|X]}{\mathbb{P}[Z=1|X]} \\ &= 0\end{aligned}$$

## Proof (Cont'd)

- Combining the last three slides, we have  $\mathbb{E}[\kappa|G, X, D, Y] = \mathbb{1}[G = c]$
- By the law of iterated expectations, we obtain

$$\begin{aligned}\mathbb{E}[\kappa f(Y, X, D)] &= \mathbb{E}[\mathbb{E}[\kappa f(Y, X, D)|G, Y, X, D]] \\ &= \mathbb{E}[f(Y, X, D)\mathbb{E}[\kappa|G, Y, X, D]] \\ &= \mathbb{E}[f(Y, X, D)\mathbb{1}[G = c]] \\ &= \mathbb{P}[G = c]\mathbb{E}[f(Y, X, D)|G = c],\end{aligned}$$

- Rearranging yields

$$\mathbb{E}[f(Y, X, D)|G = c] = \frac{1}{\mathbb{P}[G = c]}\mathbb{E}[\kappa f(Y, X, D)]$$

# References 1

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- (2003): "Semiparametric instrumental variable estimation for causal effects," *Journal of Econometrics*, 113, 231–263.
- ANDREWS, I., J. H. STOCK, AND L. SUN (2019): "Weak Instruments in Instrumental Variables Regression: Theory and Practice," *Annual Review of Economics*, 11, 727–753.
- POTERBA, J. M., S. F. VENTI, AND D. A. WISE (1994): "401(k) plans and tax-deferred savings," in *Studies in the Economics of Aging*, ed. by D. A. Wise, University of Chicago Press.
- (1995): "Do 401(k) contributions crowd out other personal savings?" *Journal of Public Economics*, 56, 1–32.