# Evaluación de Impacto: DiD

Francesco Bogliacino

#### Tabla de contenido

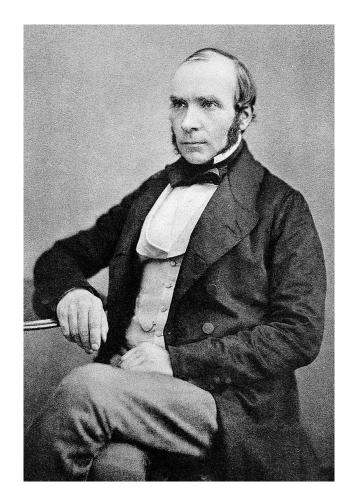
#### 1. Jon Snow

- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference

#### La historia de J Snow

- J Snow era un anestesiólogo de Londres del 1800
- A mitad del siglo XIX escribe dos obras sobre el colera
- Su tesis fundamental era: la enfermedad tiene como causa el agua

Fuente: Cunningham (2018)



Fuente: Wikipedia

#### El colera

- Mortalidad alta (50%)
- Sintomatología es muy dolorosa (diarrea y vómito) e induce deshidratación
- Vibrio cholera

- 1831-1832, 1848-1849, 1853-1854 las tres epidemias en Londres
- En esa época la teoría era la de los miasmas
- No existía teoría de bacterios y microorganismos

## Los mapas



"The most terrible outbreak of cholera which ever occurred in this kingdom, is probably that which took place in Broad Street, Golden Square, and the adjoining streets, a few weeks ago. Within two hundred and fifty yards of the spot where Cambridge Street [now Lexington St.] joins Broad Street [now Broadwick], there were upwards of five hundred fatal attacks of cholera in ten days." (Snow 1855)

## Agua vs miasmas

- Micro-organismo que entra en el intestino, se reproduce, infecta el agua e induce la expulsión
- El agua contaminada es reciclada (Londres del 800 no era la de los oligarcas rusos de hoy)
- La gente toma agua y se enferma
- Sigue el proceso

- Miasmas transmiten el "virus" (nadie sabía que era un virus en esa época)
- Más abajo, más transmisión
- Más concentración más transmisión

[la teoría es consistente con los hechos estilizados, hay correlación entre los fenómenos]

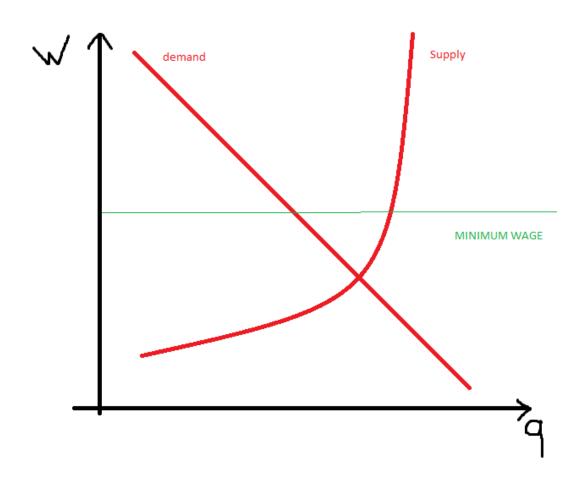
Data	Summary	Statistical Testing	Theory	Refute?	Comment
	_		Water	NO	
Albion Terrace, 17 houses & 20-25 deaths, 1849	17 houses infected, surrounding not	None	Miasma	YES	Sewage leaked into shared water supply after storm. Crucial for developing theory
Broad St – Susannah	Single Case, "Far		Water	NO	Water bottles shipped
Eley (Hampstead, 1 person)	from pump but died"	None	Miasma	YES	to Hampstead by sons
Broad St – St. James	Counterexample?		Water	NO	
vorkhouse (535 people, 5 "Close to pump but leaths) survived"	None	Miasma	YES	In-house well	
Broad St 500 residents.	Infection rates differ	Contingency	Water	NO	
categorized by drinking & illness	by pump drinking	Table	Miasma	YES	
S London 480k people,	Mortality rates differ	Diff-in-diffs,	Water	NO	Lambeth Water Co
1849 vs 1854 diff-in-diffs, aggregate sub-district	by water supply, not other characteristics	linear & count regressions, error analysis	Miasma	YES	changed to clean water 1852, ⇒ control / treatment DiD design
S London 480k people,	Mortality rates differ	RCT, Count	Water	NO	Mixing of water co
direct District / sub-district comparison,	by water supply company	regressions, detailed error	Miasma	YES	customers, control / treatment, effectively
quasi-randomized		analysis	<b>∢</b> □ Þ	4 / 17 > 4 = 1	randomized → √ = → ○ へ ○

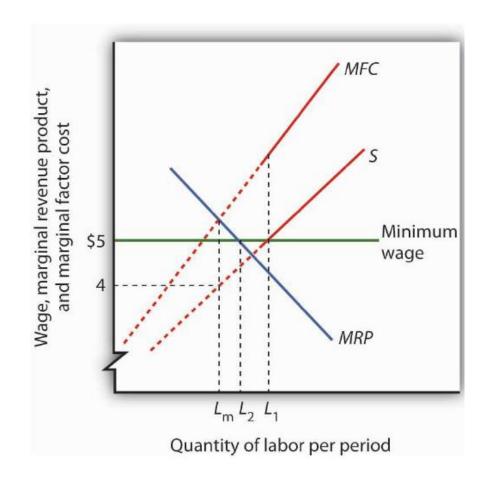
Fuente: Coleman (2019)

#### Tabla de contenido

- 1. Jon Snow
- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference

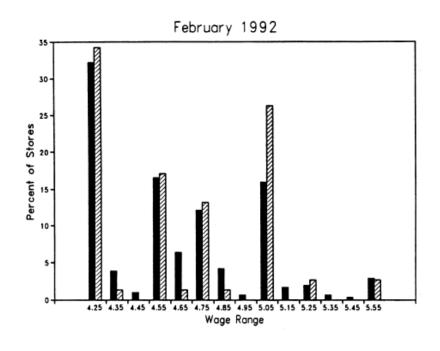
#### Predicción teórica





Source: Cunningham

- New Jersey aumenta el salario mínimo de 4.25\$ a 5.05\$, mientras que Pennsylvania lo deja a 4.25\$
- Recolección de datos orientado hacia evaluar el efecto de la política:
  - Fast food: (a) contestan encuesta; (b) tienen muchos trabajadores al mínimo;
     (c) es difícil que violen la regulación por temas de franquicia; (d) el producto es homogéneo
- Datos: feb 1992, nov 1992
  - Response rate inicial: NJ 91%; PA 72.5%
  - 90% response rate in the second wave
  - Data completed with in site visit (6 cerrados, 2 cerrado temporáneamente, 2 en renovación)-> 99.8% of second wave



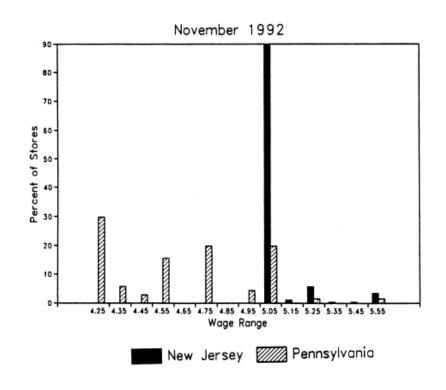


FIGURE 1. DISTRIBUTION OF STARTING WAGE RATES

TABLE 2-MEANS OF KEY VARIABLES

	Store	es in:	
Variable	NJ	PA	t <sup>a</sup>
1. Distribution of Store Types (percentages	·):		
a. Burger King	41.1	44.3	-0.5
b. KFC	20.5	15.2	1.2
c. Roy Rogers	24.8	21.5	0.6
d. Wendy's	13.6	19.0	- 1.1
e. Company-owned	34.1	35.4	-0.2
2. Means in Wave 1:			
a. FTE employment	20.4	23.3	-2.0
	(0.51)	(1.35)	
<ul> <li>b. Percentage full-time employees</li> </ul>	32.8	35.0	-0.7
	(1.3)	(2.7)	
c. Starting wage	4.61	4.63	- 0.4
	(0.02)	(0.04)	
d. Wage = \$4.25 (percentage)	30.5	32.9	-0.4
	(2.5)	(5.3)	
e. Price of full meal	3.35	3.04	4.0
	(0.04)	(0.07)	
<ol> <li>Hours open (weekday)</li> </ol>	14.4	14.5	-0.3
	(0.2)	(0.3)	
g. Recruiting bonus	23.6	29.1	-1.0
	(2.3)	(5.1)	

Table 3—Average Employment Per Store Before and After the Rise in New Jersey Minimum Wage

	Stores by state		y state	Sto	res in New Jers	ey <sup>a</sup>	Differences within NJb	
Variable	PA (i)	NJ (ii)	Difference, NJ – PA (iii)	Wage = \$4.25 (iv)	Wage = \$4.26-\$4.99 (v)	Wage ≥ \$5.00 (vi)	Low- high (vii)	Midrange- high (viii)
FTE employment before, all available observations	23.33 (1.35)	20.44 (0.51)	-2.89 (1.44)	19.56 (0.77)	20.08 (0.84)	22.25 (1.14)	-2.69 (1.37)	-2.17 (1.41)
<ol><li>FTE employment after, all available observations</li></ol>	21.17 (0.94)	21.03 (0.52)	-0.14 (1.07)	20.88 (1.01)	20.96 (0.76)	20.21 (1.03)	0.67 (1.44)	0.75 (1.27)
<ol> <li>Change in mean FTE employment</li> </ol>	-2.16 (1.25)	0.59 (0.54)	2.76 (1.36)	1.32 (0.95)	0.87 (0.84)	-2.04 (1.14)	3.36 (1.48)	2.91 (1.41)
<ol> <li>Change in mean FTE employment, balanced sample of stores<sup>c</sup></li> </ol>	-2.28 (1.25)	0.47 (0.48)	2.75 (1.34)	1.21 (0.82)	0.71 (0.69)	-2.16 (1.01)	3.36 (1.30)	2.87 (1.22)
<ol> <li>Change in mean FTE employment, setting FTE at temporarily closed stores to 0<sup>d</sup></li> </ol>	-2.28 (1.25)	0.23 (0.49)	2.51 (1.35)	0.90 (0.87)	0.49 (0.69)	-2.39 (1.02)	3.29 (1.34)	2.88 (1.23)

TABLE 4—REDUCED-FORM MODELS FOR CHANGE IN EMPLOYMENT

	Model							
Independent variable	(i)	(ii)	(iii)	(iv)	(v)			
New Jersey dummy	2.33 (1.19)	2.30 (1.20)	_	_	_			
2. Initial wage gap <sup>a</sup>	_	-	15.65 (6.08)	14.92 (6.21)	11.91 (7.39)			
<ol> <li>Controls for chain and ownership<sup>b</sup></li> </ol>	no	yes	no	yes	yes			
<ol> <li>Controls for region<sup>c</sup></li> </ol>	no	no	no	no	yes			
<ol><li>Standard error of regression</li></ol>	8.79	8.78	8.76	8.76	8.75			
<ol> <li>Probability value for controls<sup>d</sup></li> </ol>		0.34	_	0.44	0.40			

TABLE 6-EFFECTS OF MINIMUM-WAGE INCREASE ON OTHER OUTCOMES

	Mean o	hange in	outcome	Regression of change in outcome variable on:			
Outcome measure	NJ	PA	NJ-PA	NJ dummy	Wage gap <sup>a</sup>	Wage gap <sup>b</sup>	
	(i)	(ii)	(iii)	(iv)	(v)	(vi)	
Store Characteristics:							
1. Fraction full-time workers <sup>c</sup> (percentage)	2.64	-4.65	7.29	7.30	33.64	20.28	
	(1.71)	(3.80)	(4.17)	(3.96)	(20.95)	(24.34)	
2. Number of hours open per weekday	-0.00	0.11	-0.11	-0.11	-0.24	0.04	
	(0.06)	(0.08)	(0.10)	(0.12)	(0.65)	(0.76)	
3. Number of cash registers	-0.04	0.13	-0.17	-0.18	-0.31	0.29	
	(0.04)	(0.10)	(0.11)	(0.10)	(0.53)	(0.62)	
<ol> <li>Number of cash registers open</li></ol>	-0.03	-0.20 (0.08)	0.17	0.17	0.15	-0.47	
at 11:00 а.м.	(0.05)		(0.10)	(0.12)	(0.62)	(0.74)	
Employee Meal Programs:							
5. Low-price meal program (percentage)	-4.67	-1.28	-3.39	-2.01	-30.31	-33.15	
	(2.65)	(3.86)	(4.68)	(5.63)	(29.80)	(35.04)	
6. Free meal program (percentage)	8.41	6.41	2.00	0.49	29.90	36.91	
	(2.17)	(3.33)	(3.97)	(4.50)	(23.75)	(27.90)	
<ol><li>Combination of low-price and free meals (percentage)</li></ol>	-4.04	-5.13	1.09	1.20	-11.87	-19.19	
	(1.98)	(3.11)	(3.69)	(4.32)	(22.87)	(26.81)	
Wage Profile:							
8. Time to first raise (weeks)	3.77	1.26	2.51	2.21	4.02	-5.10	
	(0.89)	(1.97)	(2.16)	(2.03)	(10.81)	(12.74)	
9. Usual amount of first raise (cents)	-0.01	-0.02	0.01	0.01	0.03	0.03	
	(0.01)	(0.02)	(0.02)	(0.02)	(0.11)	(0.11)	
<ol><li>Slope of wage profile (percent</li></ol>	-0.10	-0.11	0.01	0.01	-0.09	-0.08	
per week)	(0.04)	(0.09)	(0.10)	(0.10)	(0.56)	(0.57)	

TABLE 7-REDUCED-FORM MODELS FOR CHANGE IN THE PRICE OF A FULL MEAL

	Dependent variable: change in the log price of a full meal							
Independent variable	(i)	(ii)	(iii)	(iv)	(v)			
New Jersey dummy	0.033 (0.014)	0.037 (0.014)	_	_	_			
2. Initial wage gap <sup>a</sup>	_	_	0.077 (0.075)	0.146 (0.074)	0.063 (0.089)			
<ol> <li>Controls for chain and<sup>b</sup> ownership</li> </ol>	no	yes	no	yes	yes			
<ol> <li>Controls for region<sup>c</sup></li> </ol>	no	no	no	no	yes			
5. Standard error of regression	0.101	0.097	0.102	0.098	0.097			

- Ejemplo de studio que es transparente, bien diseñado desde la recolección
- Todo el ejercicio es informado por la teoría
- Fue muy influyente y generó debats muy encendidos

#### Tabla de contenido

- 1. Jon Snow
- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference

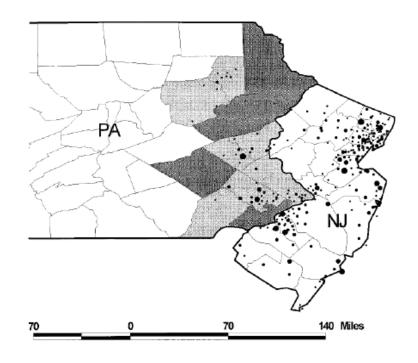
## El salario mínimo y el empleo

- New Jersey aumenta el salario mínimo de 4.25\$ a 5.05\$, mientras que Pennsylvania lo deja a 4.25\$
- Para el estado s al tiempo t el empleo con salario mínimo (k) bajo/alto [k=0,1] es

$$E[Y_{ist}^{k}|s,t] = \gamma_{s} + \tau_{t}$$

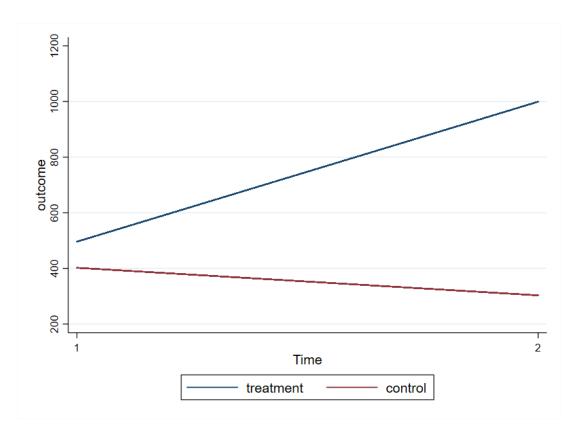
γ

$$E[Y_{ist}^{1} - Y_{ist}^{0} | s, t] = \delta$$



## The switching regression

$$\begin{split} Y_{it} &= D_{i}^{treat} \left[ D_{i}^{t1} (Y_{i,t1}^{1}) + (1 - D_{i}^{t1}) (Y_{i,t0}^{1}) \right] \\ &+ (1 - D_{i}^{treat}) \left[ D_{i}^{t1} (Y_{i,t1}^{0}) + (1 - D_{i}^{t1}) (Y_{i,t0}^{0}) \right] = \\ &+ (1 - D_{i}^{t1}) (Y_{i,t0}^{0}) \right] = \\ &Y_{i,t0}^{0} + D_{i}^{t1} (Y_{i,t1}^{0} - Y_{i,t0}^{0}) + \\ &+ D_{i}^{treat} (Y_{i,t0}^{1} - Y_{i,t0}^{0}) + \\ &+ D_{i}^{treat} D_{i}^{t1} \left[ (Y_{i,t1}^{1} - Y_{i,t0}^{1}) - (Y_{i,t0}^{0} - Y_{i,t0}^{0}) \right] \end{split}$$



## La lógica de Dif in Dif

$$E[Y_{ist}^{k}|s,t] = \gamma_{s} + \tau_{t}$$

$$Y$$

$$E[Y_{ist}^{1} - Y_{ist}^{0}|s,t] = \delta$$

$$Y_{ist} = \gamma_{s} + \tau_{t} + \delta D_{st} + \varepsilon_{ist}$$

$$E[Y_{ist}|NJ,post] = \gamma_{NJ} + \tau_{t} + \delta$$

$$E[Y_{ist}|NJ,pre] = \gamma_{NJ}$$

$$E[Y_{ist}|PA,post] = \gamma_{PA} + \tau_{t} + \delta$$

$$E[Y_{ist}|PA,pre] = \gamma_{PA}$$

$$E[Y_{ist}|NJ,post] - E[Y_{ist}|NJ,pre]$$

$$= \tau_t + \delta$$
No as suficients

No es suficiente

$$E[Y_{ist}|PA,post] - E[Y_{ist}|PA,pre] = \tau_t$$

No es suficiente

$$E[Y_{ist}|NJ,post] - E[Y_{ist}|NJ,pre]$$

 $-|E[Y_{ist}|PA, post]$ 

$$-E[Y_{ist}|PA,pre] = \delta$$

Si las tendencias son paralelas

## Potential Outcome & Switching Equation

$$\begin{aligned} & \left[ E \left( Y_{T,post}^{1} - Y_{T,pre}^{1} \right) - E \left( Y_{C,post}^{0} - Y_{C,pre}^{0} \right) \right] = \\ & \left[ E \left( Y_{T,post}^{1} - Y_{T,pre}^{1} \right) - E \left( Y_{C,post}^{0} - Y_{C,pre}^{0} \right) \right] + E \left[ Y_{T,post}^{0} \right] - E \left[ Y_{T,post}^{0} \right] = \end{aligned}$$

$$E\left(Y_{T,post}^{1}-Y_{T,post}^{0}\right)=$$

$$E\left[Y_{T,post}^{0}-Y_{T,pre}^{1}\right]-E\left(Y_{C,post}^{0}-Y_{C,pre}^{0}\right)$$

La diferencia entre "trends": si las tendencias son paralelas es igual a cero

#### Parallel trends

- El DiD se puede usar con *repeated cross section*, por eso es tal vez el más común de los diseños de investigación
- Sin parallel trends estoy violando identif
- Si cambia la composición de los grupos puede ser muy común la variación de tendencias paralelas

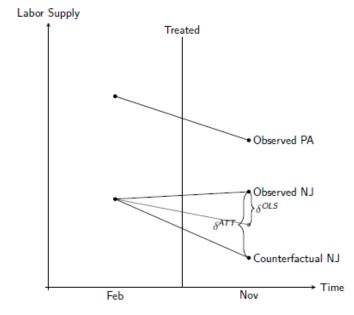
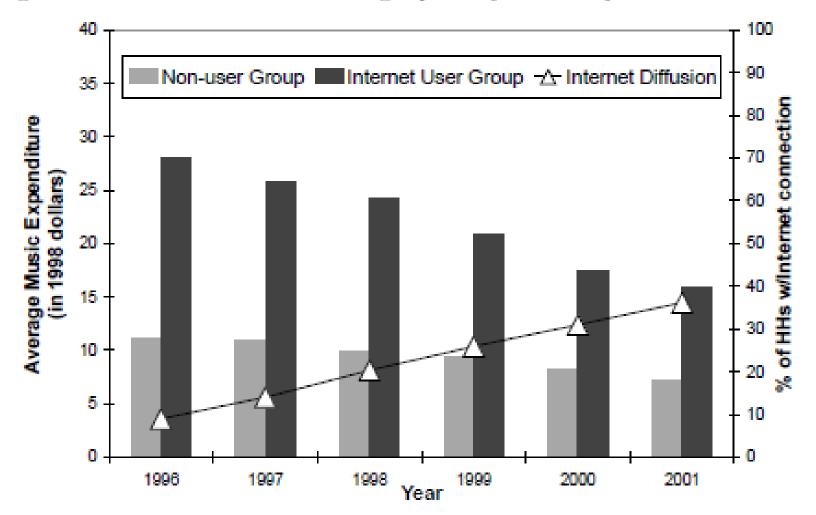


Figure: DD regression diagram without parallel trends

Table 1: Descriptive Statistics for Internet User and Non-user Groups\*

Year	1997		1998	1998		
	Internet User	Non-user	Internet User	Non-user	Internet User	Non-user
Average Expenditure						
Recorded Music	\$25.73	\$10.90	\$24.18	\$9.97	\$20.92	\$9.37
Entertainment	\$195.03	\$96.71	\$193.38	\$84.92	\$182.42	\$80.19
Zero Expenditure						
Recorded Music	.56	.79	.60	.80	.64	.81
Entertainment	.08	.32	.09	.35	.14	.39
Demographies						
Age	40.2	49.0	42.3	49.0	44.1	49.4
Income	\$52,887	\$30,459	\$51,996	\$28,169	\$49,970	\$26,649
High School Grad.	.18	.31	.17	.32	.21	.32
Some College	.37	.28	.35	.27	.34	.27
College Grad.	.43	.21	.45	.21	.42	.20
Manager	.16	.08	.16	.08	.14	.08

Figure 1: Internet Diffusion and Average Quarterly Music Expenditure in the CEX



#### Tabla de contenido

- 1. Jon Snow
- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference

649	325	0	NJ	19.16265	1	0	0
650	325	1	NJ	19.83828	1	1	1
651	326	0	NJ	21.34171	1	0	$\setminus$ $\bigcirc$
652	326	_ 1	NJ	20.64373	1	1	ノしュノ
653	327	0	NJ	20.80775	1	0	0
654	327	1	NJ	20.96773	1	1	1
655	328	0	NJ	19.56505	1	0	0
656	328	1	NJ	20.39766	1	1	1
657	329	0	NJ	20.68392	1	0	0
658	329	1	NJ	22.0185	1	1	1
659	330	0	NJ	20.36015	1	0	0
660	330	1	NJ	21.88702	1	1	1
661	331	0	NJ	21.43298	1	0	0
662	331	1	NJ	21.13688	1	1	1
663	332	0	PA	23.62994	0	0	0
664	332		PA	19.59755	0	1	0
665	333	0	PA	23.68527	0	0	0
666	333	1	PA	20.46726	0	1	0
667	334	0	PA	21.45385	0	0	0
668	334	1	PA	22.15355	0	1	0

### La estructura de la base de datos

#### **OLS**

$$y_{ist} = \alpha + \beta N J_i + \gamma d_t + \delta (N J_i d_t) + \varepsilon_{ist}$$

- NJ es igual a uno para NJ y 0 para PA
- Bajo parallel trends
  - Para PA, pre=  $\alpha$
  - Para PA, post=  $\alpha + \gamma$
  - Para NJ, pre=  $\alpha + \beta$
  - Para NJ, post=  $\alpha + \beta + \gamma + \delta$

\_

. reg empl NJ expost NJ expost, rob

Linear regression	Number of obs	=	820
	F(3, 816)	=	164.53
	Prob > F	=	0.0000
	R-squared	=	0.5734
	Root MSE	=	. 6812

empl	Coef.	Robust Std. Err.	t	P> t	[95% Conf.	Interval
	Coer.	oca. EII.		1/ 0	[55% CONT.	incervar.
NJ	-2.822551	.159201	-17.73	0.000	-3.135043	-2.510059
expost	-2.252137	.1825651	-12.34	0.000	-2.61049	-1.893785
NJ_expost	2.861982	.1868274	15.32	0.000	2.495263	3.22870
_cons	23.23443	.1566275	148.34	0.000	22.92699	23.5418

• DiD Estimate: (NJ, post-NJ, pre)-(NJ, post-NJ, pre)=  $\alpha + \beta + \gamma + \delta - \alpha - \beta - (\alpha + \gamma - \alpha) = \delta$ 

#### **TWFE**

#### $y_{ist} = \alpha + \gamma d_t + \delta(NJ_i d_t) + \mu_i + \varepsilon_{ist}$

. xtreg empl expost NJ\_expost, fe vce(rob)

Fixed-effects (within) regression	Number of obs $=$	820
Group variable: iid	Number of groups =	410
R-sq:	Obs per group:	
within $= 0.5741$	min =	2
between = 0.5726	avg =	2.0
overall = 0.0001	max =	2
	F(2,409) =	190.01
$corr(u_i, Xb) = -0.6501$	Prob > F =	0.0000

(Std. Err. adjusted for 410 clusters in iid)

empl	Coef.	Robust Std. Err.	t	P> t	[95% Conf.	Interval]
expost NJ_expost _cons	-2.252137 2.861982 20.95573	.186049 .1902816 .0241002	-12.11 15.04 869.53	0.000 0.000 0.000	-2.617869 2.48793 20.90836	-1.886406 3.236034 21.00311
sigma_u sigma_e rho	1.2114728 .69012486 .75499622	(fraction	of varia	nce due t	:o u_i)	

#### Tabla de contenido

- 1. Jon Snow
- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference

#### Los diseños DiD-> State level variation



## BDM

TABLE II
DD REJECTION RATES FOR PLACEBO LAWS

A. CPS DATA				
			Rejection rate	
Data	$\hat{\rho}_1,\hat{\rho}_2,\hat{\rho}_3$	Modifications	No effect	2% effect
1) CPS micro, log wage 2) CPS micro, log wage 3) CPS agg, log wage 4) CPS agg, log wage 5) CPS agg, log wage 6) CPS agg,	.509, .440, .332 .509, .440, .332 .509, .440, .332 .470, .418, .367	Cluster at state- year level  Sampling w/replacement Serially uncorrelated laws	.675 (.027) .44 (.029) .435 (.029) .49 (.025) .05 (.011)	.855 (.020) .74 (.025) .72 (.026) .663 (.024) .988 (.006)
employment 7) CPS agg, hours worked 8) CPS agg, changes in log wage	.151, .114, .063		(.025) .265 (.022)	(.016) .280 (.022) .978 (.007)

#### B. MONTE CARLO SIMULATIONS WITH SAMPLING FROM AR(1) DISTRIBUTION

Data			Rejection rate	
	ρ	Modifications	No effect	2% effect
9) AR(1)	.8		.373	.725
			(.028)	(.026)
10) AR(1)	0		.053	.783
			(.013)	(.024)
11) AR(1)	.2		.123	.738
			(.019)	(.025)
12) AR(1)	.4		.19	.713
			(.023)	(.026)
13) AR(1)	.6		.333	.700
			(.027)	(.026)
14) AR(1)	4		.008	.7
			(.005)	(.026)

#### **BDM**

• <DiD.do>

- BDM no sugieren modelar el término de error
- Sugieren:
  - Hacer block randomization a nivel de estado
  - Clustering a nivel de estado
  - Collapsar la info a nivel de estado antes y después y correr la regresión

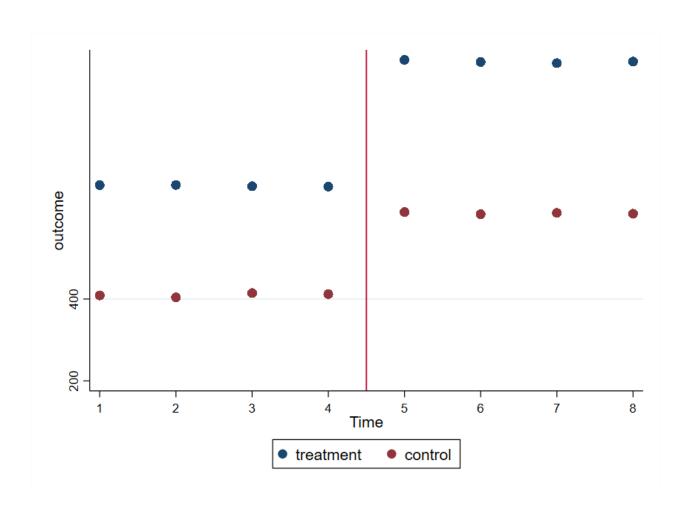
#### Tabla de contenido

- 1. Jon Snow
- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference

## Ejemplo

- 50 estados
- La ley se pasa entre t=4 y t=5
- T=1, ..., 8
- Queremos ver el impacto sobre ingresos

## (1) Plot the data to see parallel trend

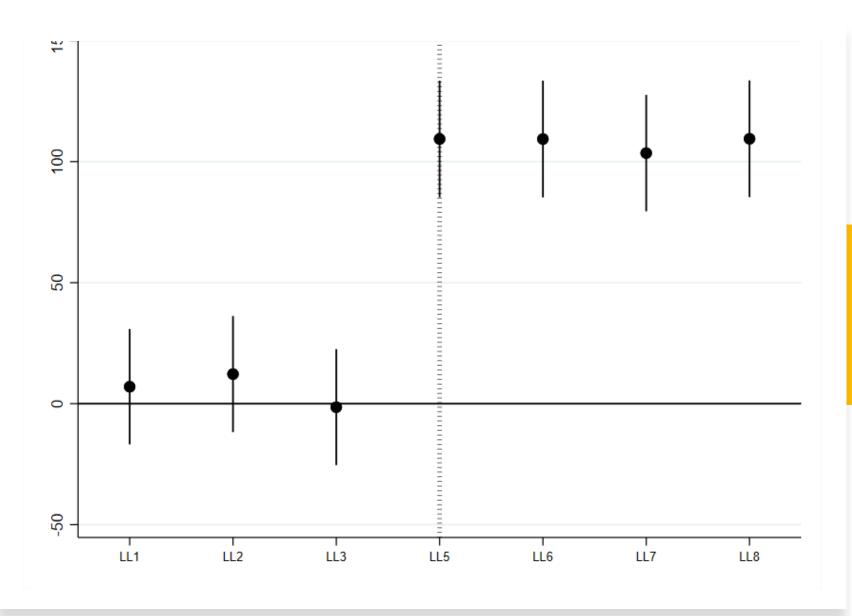


## (2) Baseline Results

	(1)	(2)	(3)
Dependent Variable: Outcome	OLS	OLS	FE
DiD	96.37***	96.37***	96.37***
	(7.38)	(6.05)	(5.10)
Constant	366.67***	379.71***	514.68***
	(3.73)	(11.50)	(2.77)
Observations	80,000	80,000	80,000
R-squared	0.14	0.42	0.18
Treatment Dummy	Yes		
Expost Dummy	Yes		
Standard errors	Rob	Rob	Cluster(iid)
Year FE		Yes	Yes
State FE		Yes	
Number of iid			10,000
Individual FE			Yes

## (3) BDM

	(1)	(2)	(3)
Outcome	FE	FE	OLS
D:D	05 24***	05 24***	05 24***
DiD	95.34***	95.34***	95.34***
	(4.82)	(4.02)	(0.24)
Constant	548.88***	548.88***	289.77***
	(2.61)	(4.38)	(0.43)
Observations	80,000	80,000	20,000
R-squared	0.18	0.18	1.00
Number of iid	10,000	10,000	
Year FE	Yes	Yes	Yes
Individual FE	Yes	Yes	
Standard errors	Cluster(state)	<b>Block Boots</b>	Averaged Pre and Post
State FE			Yes



$$y_{ist} = \sum_{\tau=-2}^{-q} \gamma_{\tau} D_i + \sum_{\tau=0}^{m} \delta_{\tau} D_i$$
$$\mu_s + \lambda_t + \zeta_i + \varepsilon_{ist}$$

## (4) Event study

## (5) Placebo

	(1)	(2)
	FE	FE
	Outcome	Placebo Outcome
DiD (placebo)	2.13	
	(10.29)	
DiD		-6.70
		(5.73)
Constant	525.01***	1,624.22***
	(4.60)	(3.80)
Observations	80,000	80,000
R-squared	0.17	0.21
Number of iid	10,000	10,000
Year FE	Yes	Yes
Individual FE	Yes	Yes
Standard errors	Cluster state	Cluster state

#### Tabla de contenido

- 1. Jon Snow
- 2. Salario mínimo y empleo
- 3. Identificación en DiD
- 4. Estimación en DiD
- 5. Errores estándares
- 6. Presentación
- 7. Triple Difference