

Introduction to Difference in Differences Using Stata

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- You can download all of the slides, datasets and do-files here:

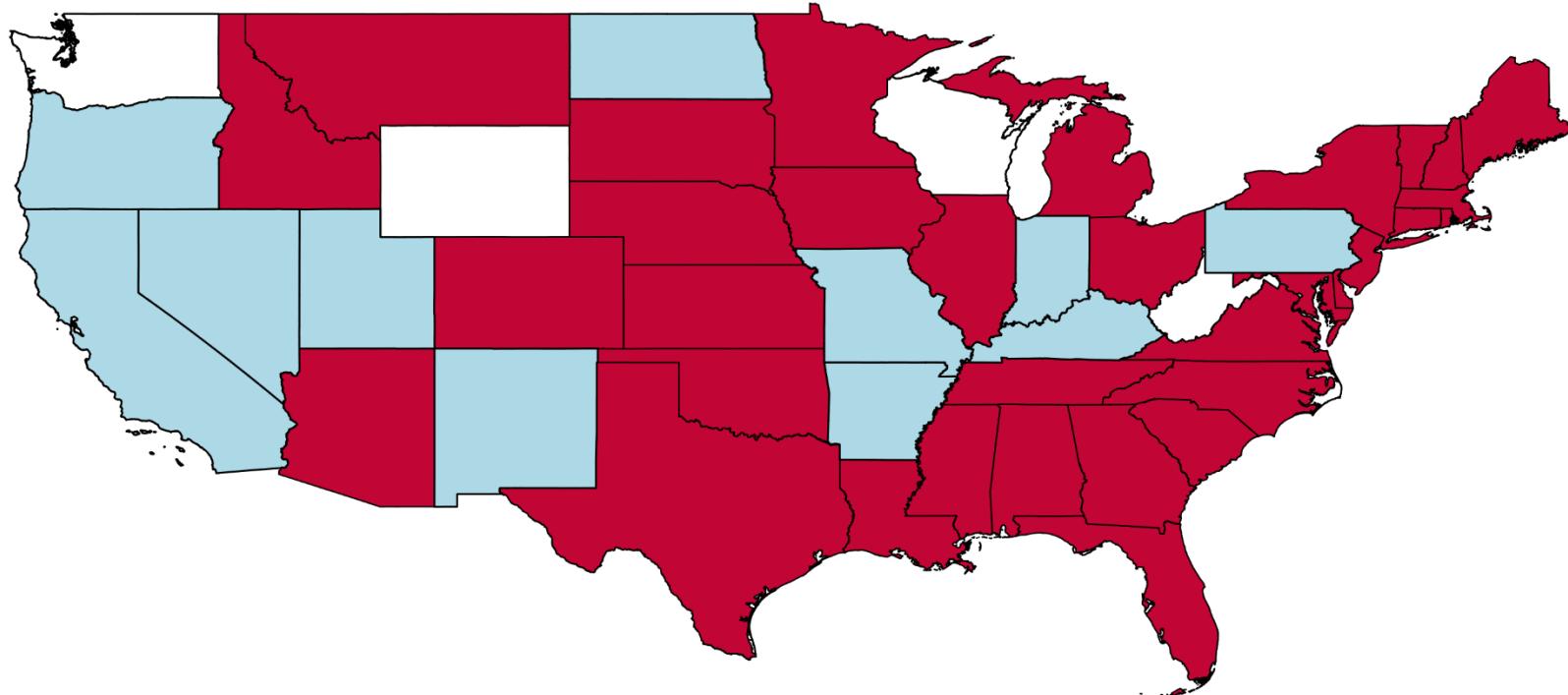
<https://tinyurl.com/StataDID>

Outline

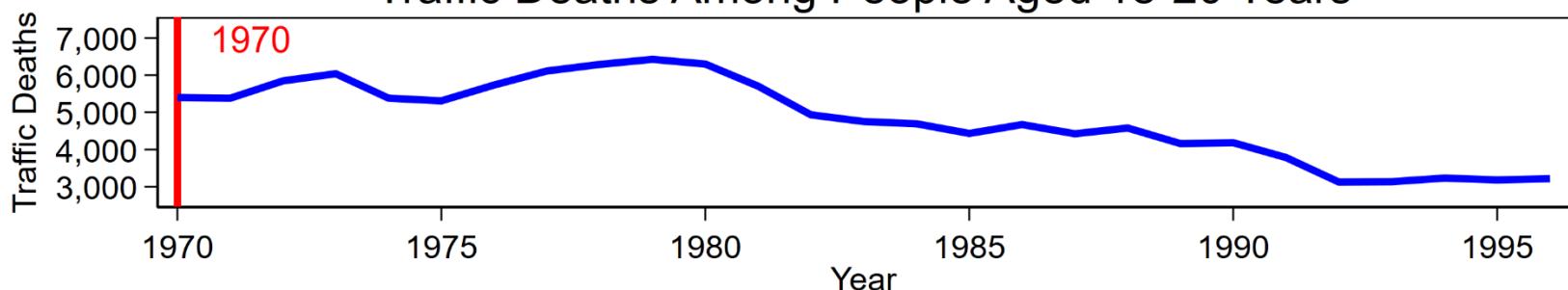
- The Question
- An Intuitive Introduction
- Two Period, Two Groups Model
- Repeated Cross-Sectional Panel Data
- Longitudinal Panel Data
- More information

States With MLDA Less than 21 (red)

Year = 1970

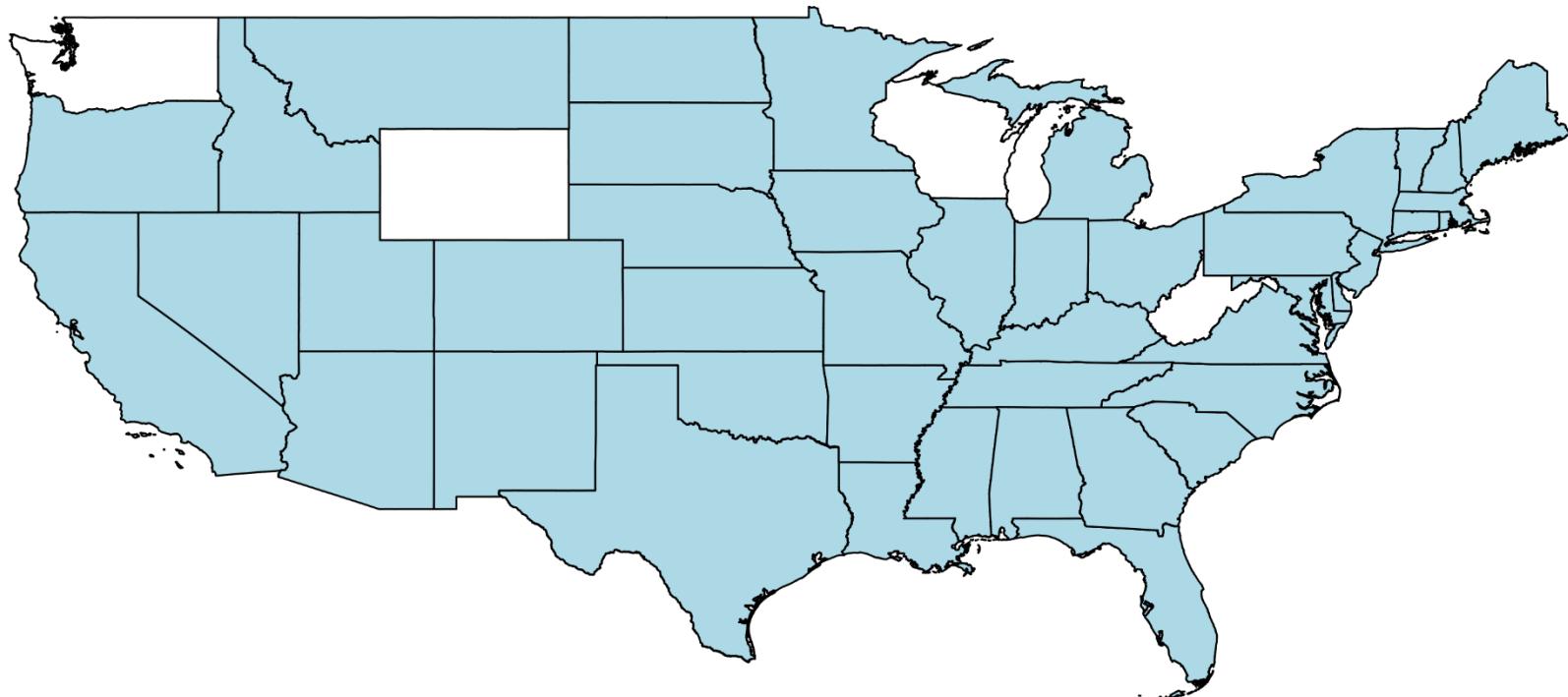


Traffic Deaths Among People Aged 18-20 Years

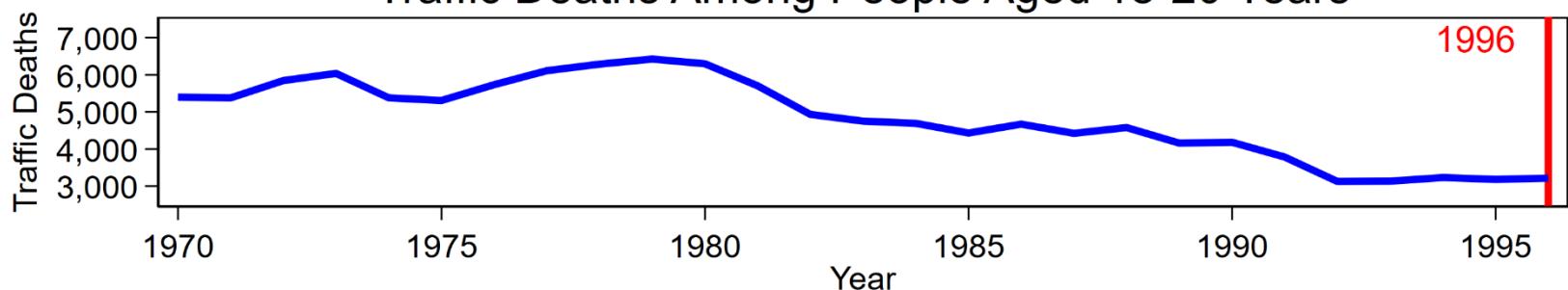


States With MLDA Less than 21 (red)

Year = 1996



Traffic Deaths Among People Aged 18-20 Years



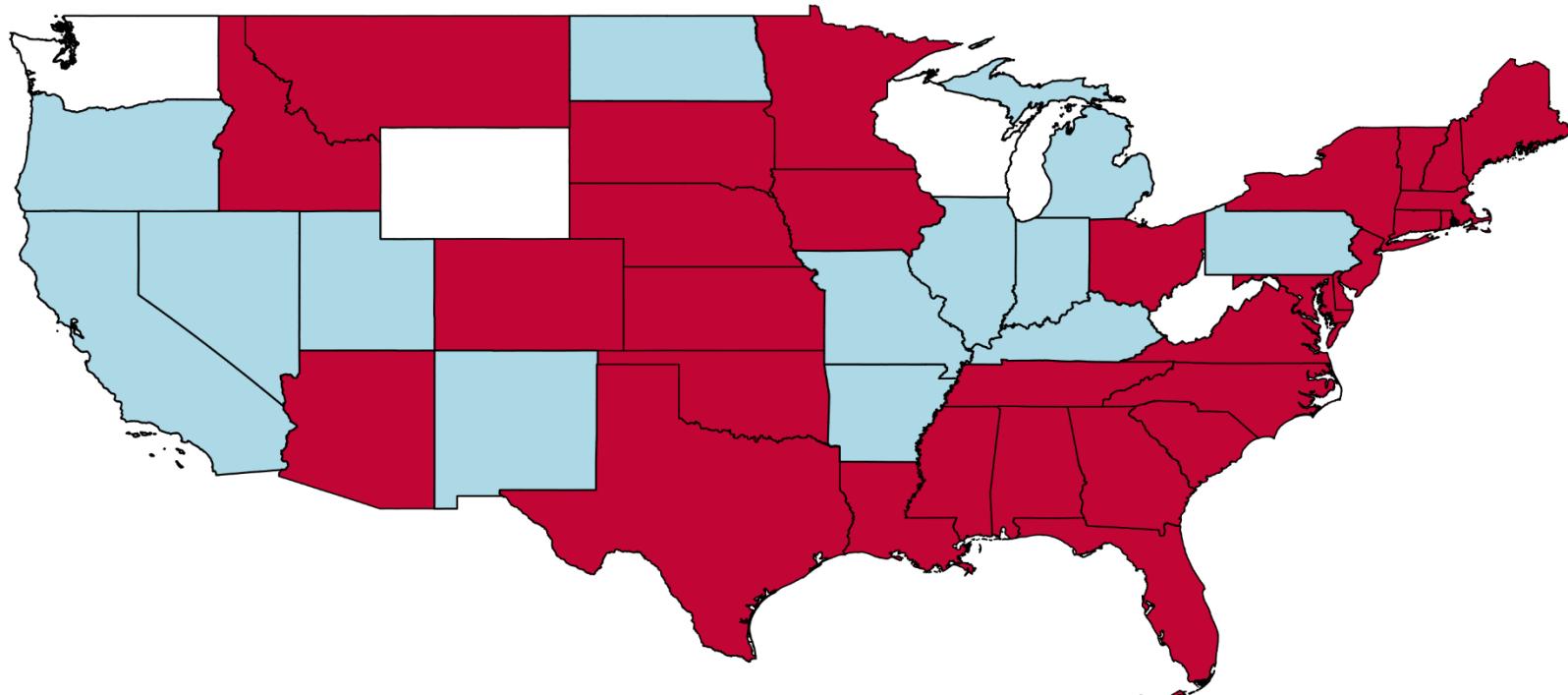


Data Analysis and Statistical Software

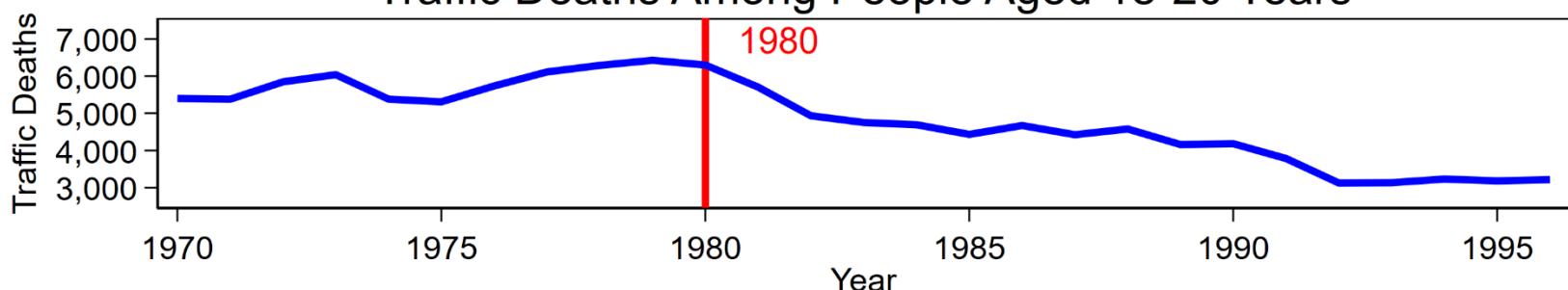


States With MLDA Less than 21 (red)

Year = 1980



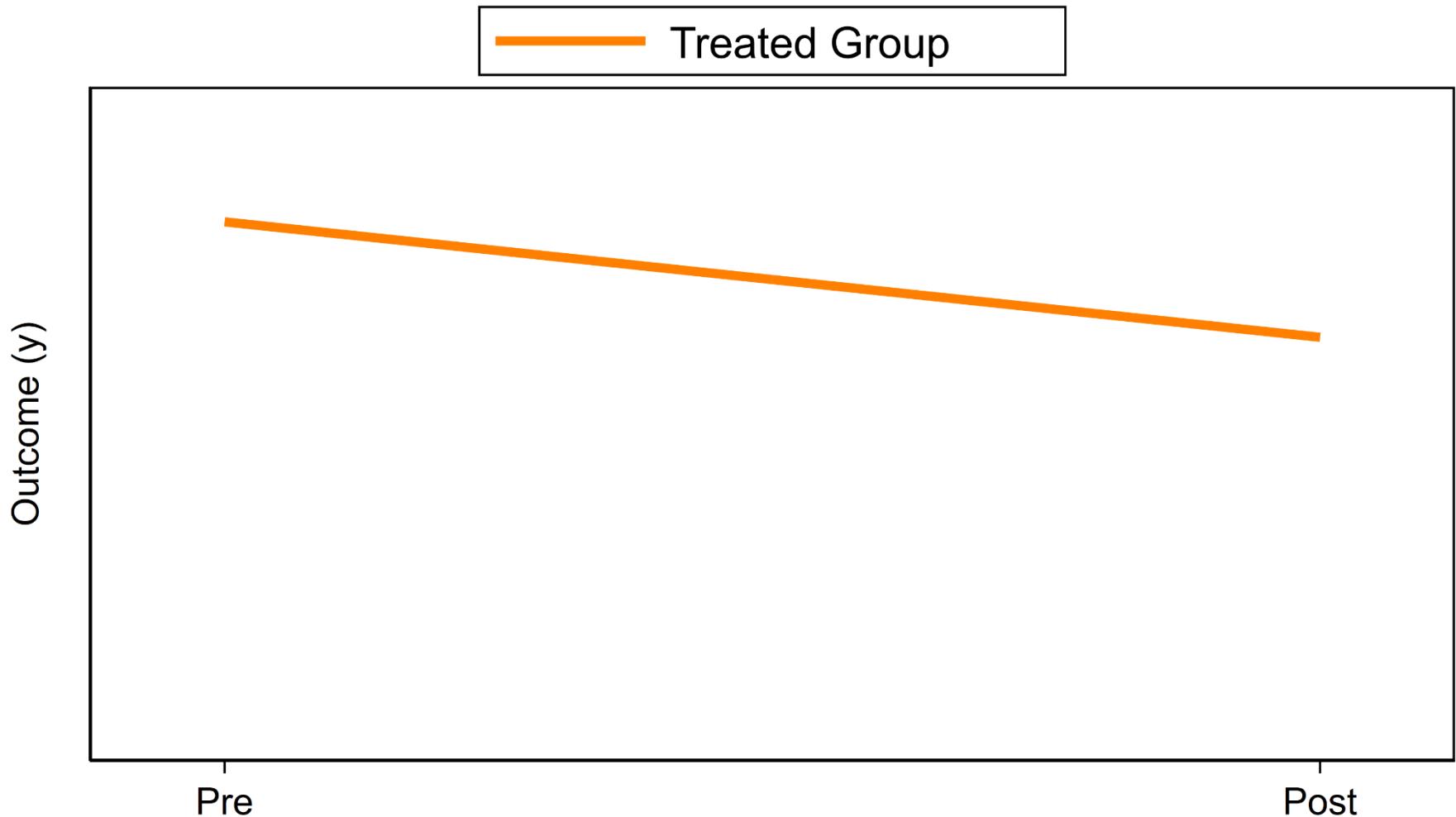
Traffic Deaths Among People Aged 18-20 Years



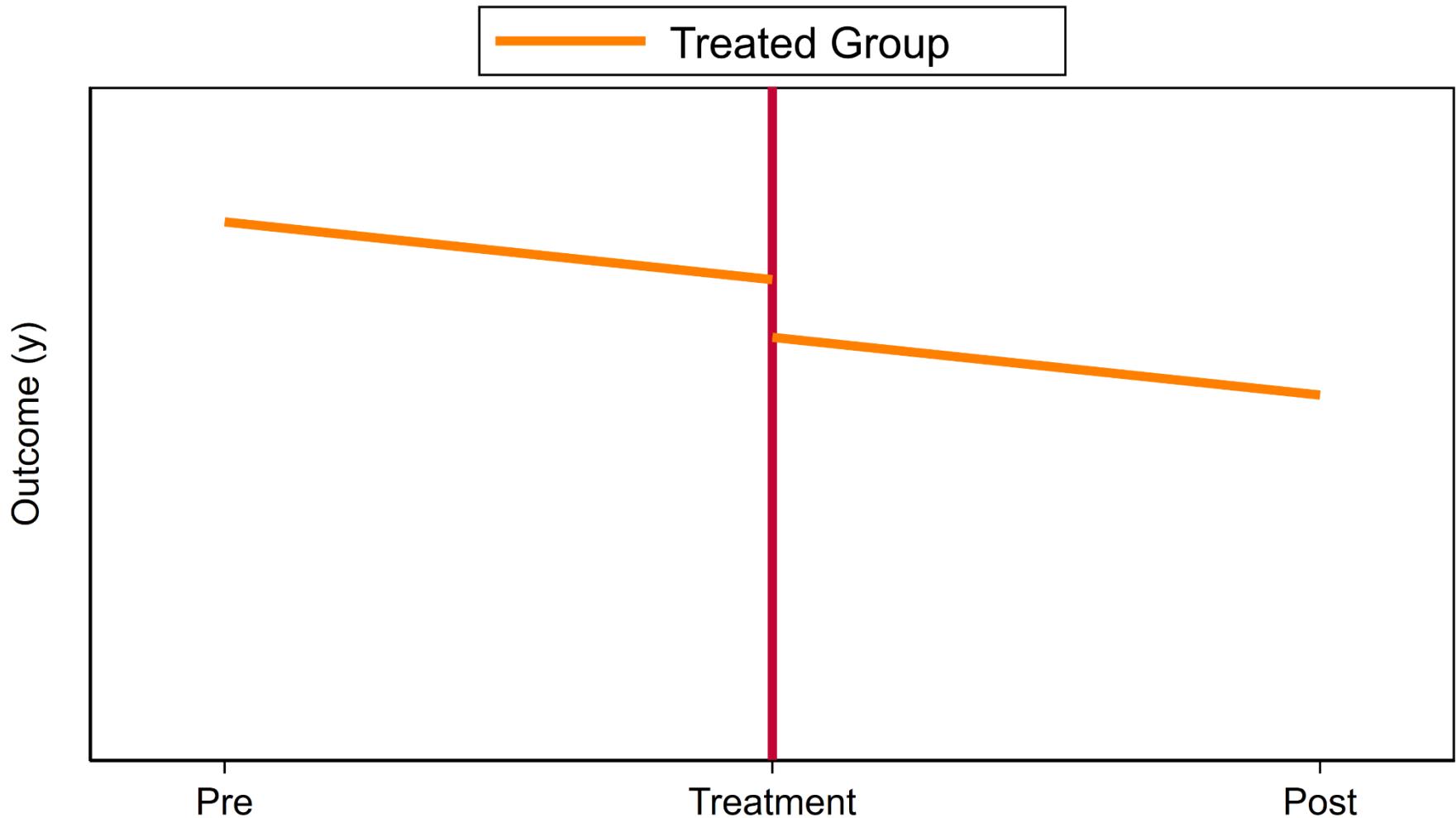
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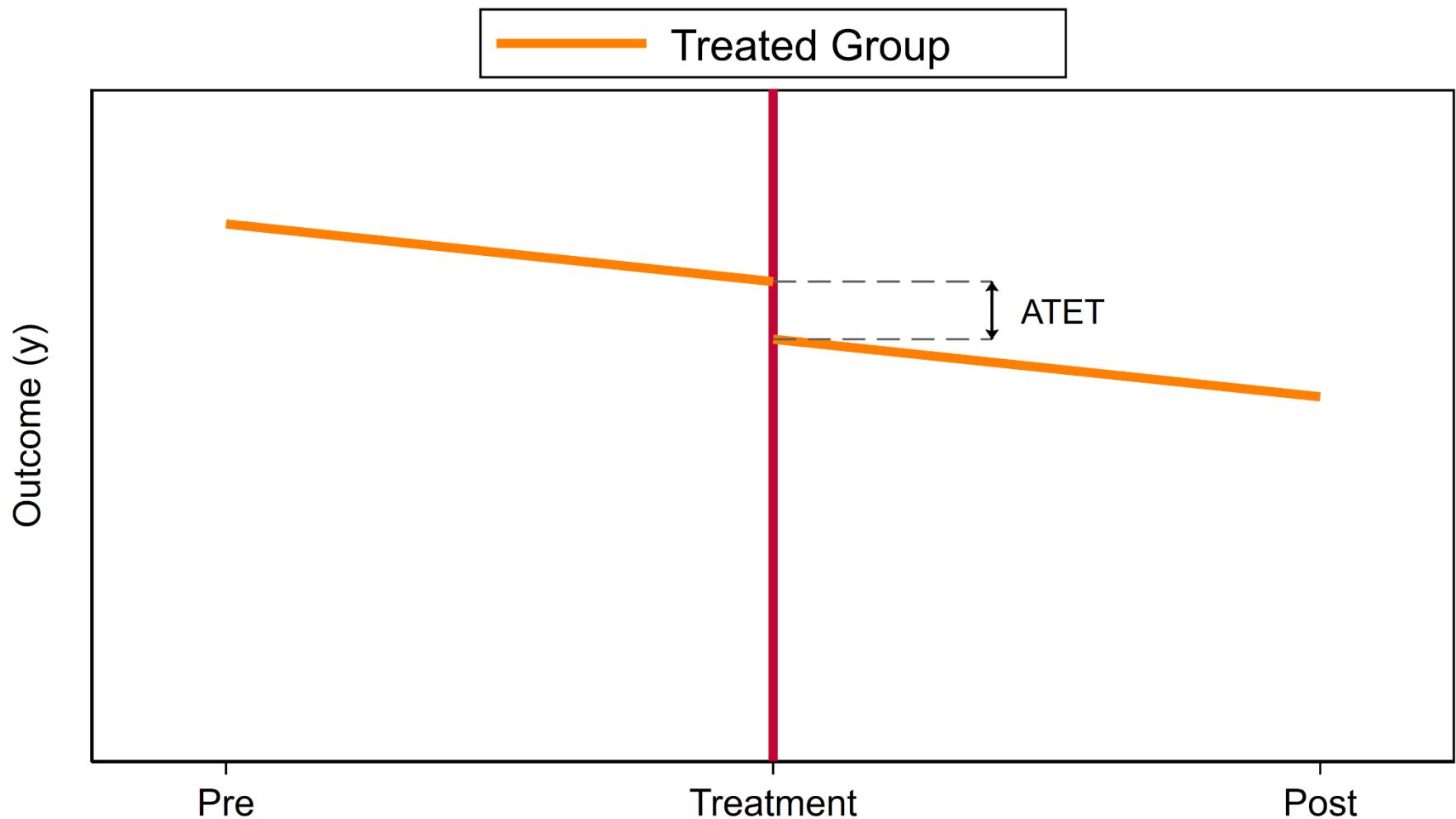
A Conceptual Introduction



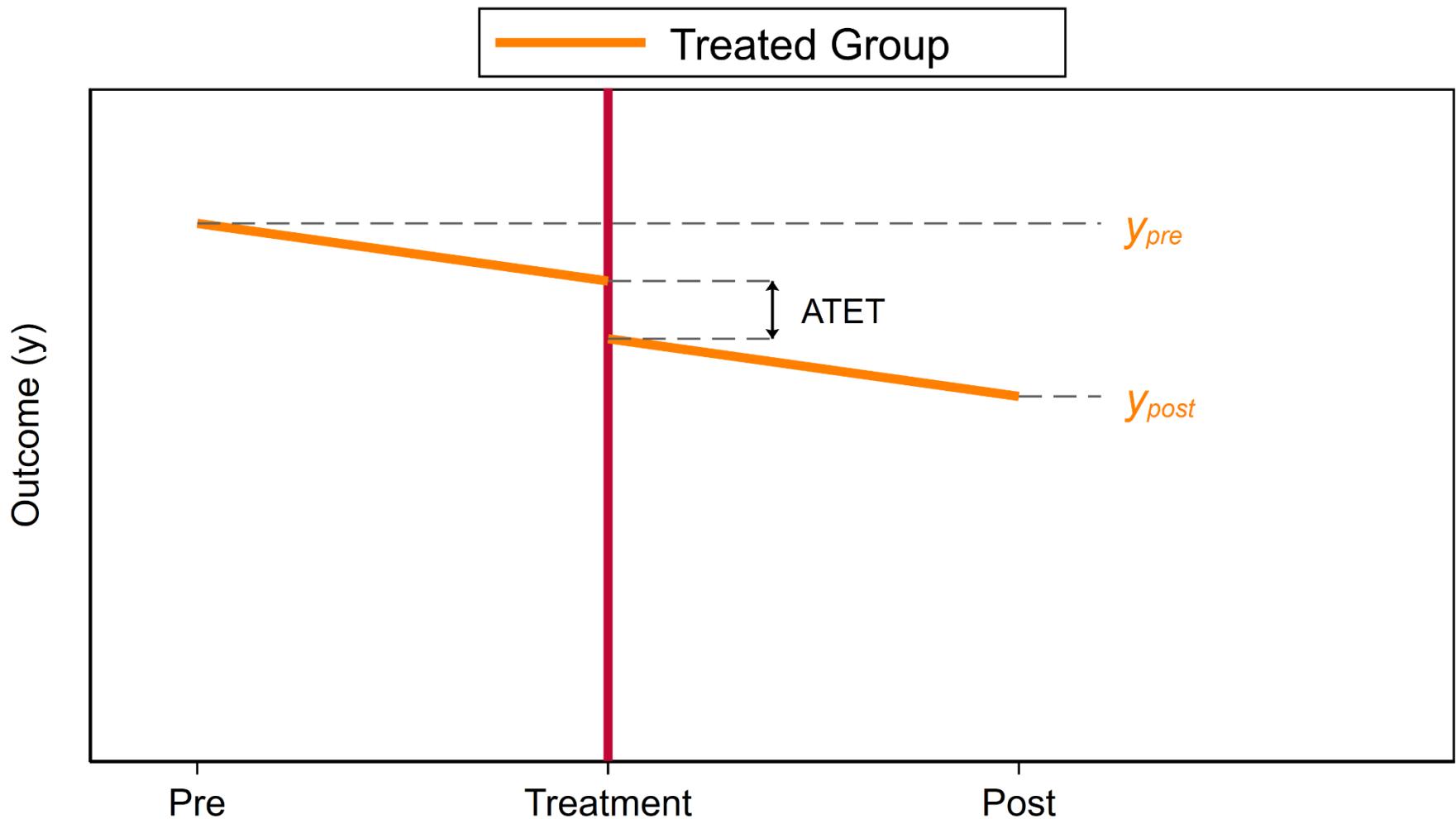
A Conceptual Introduction



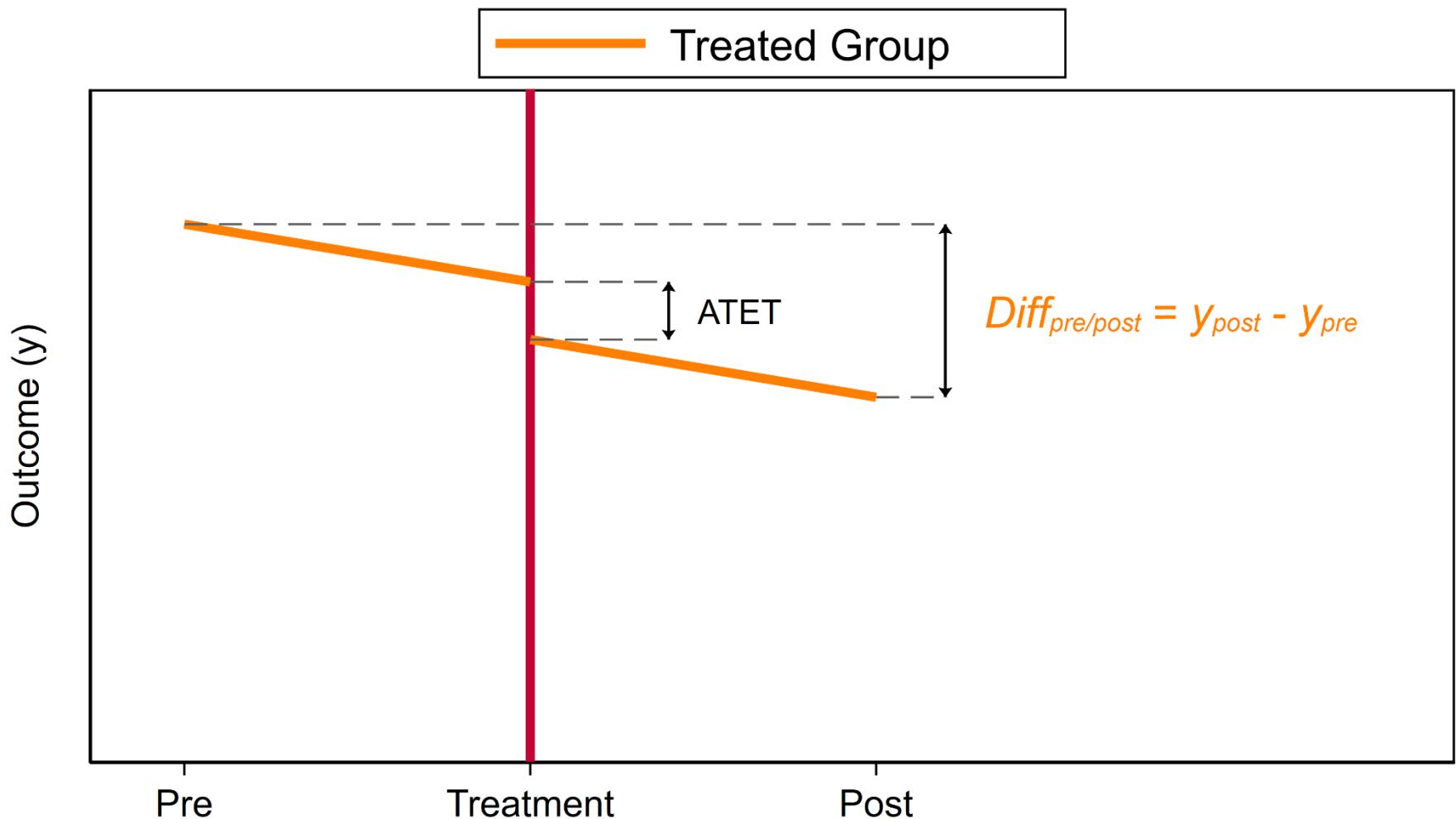
A Conceptual Introduction



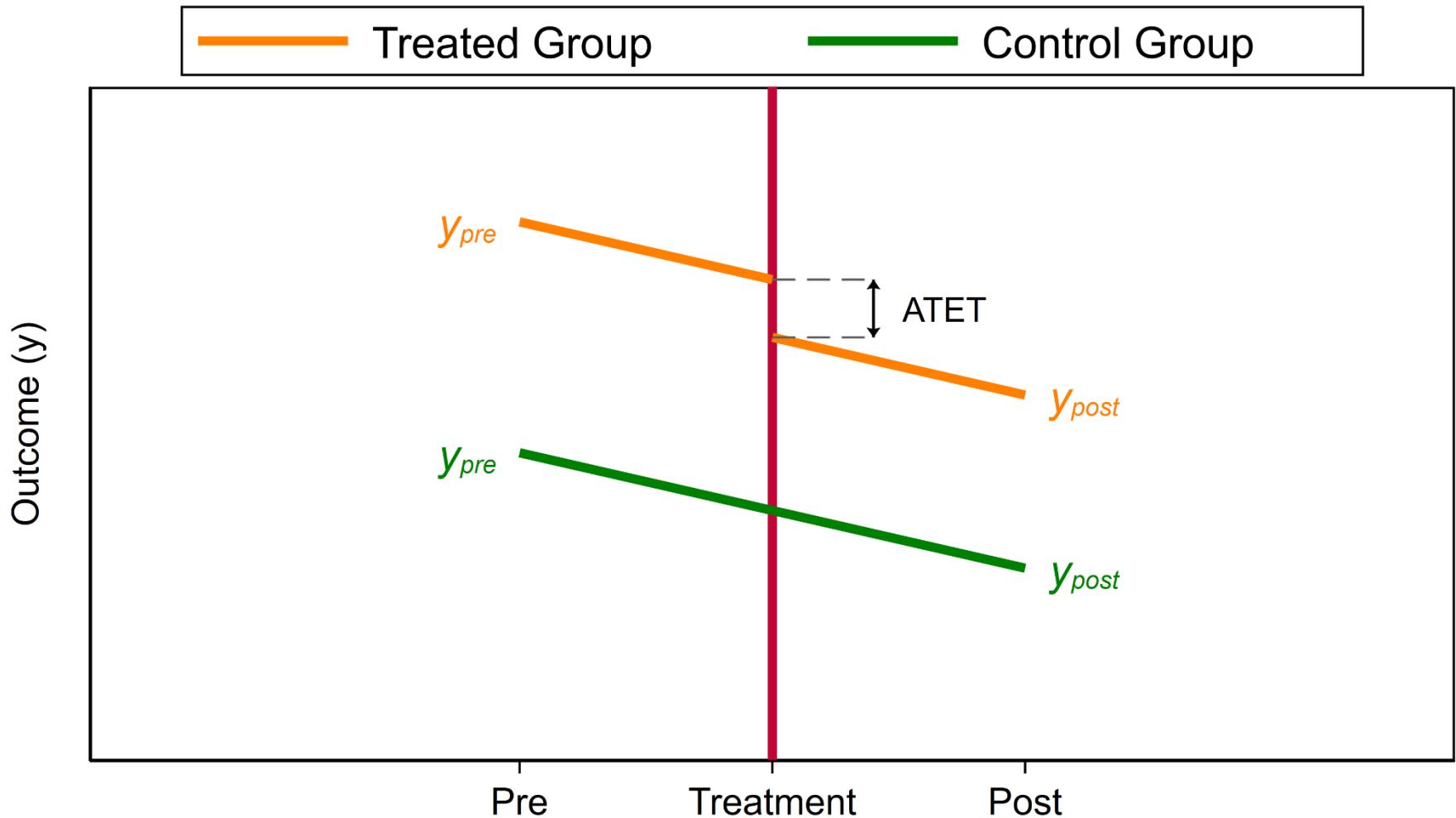
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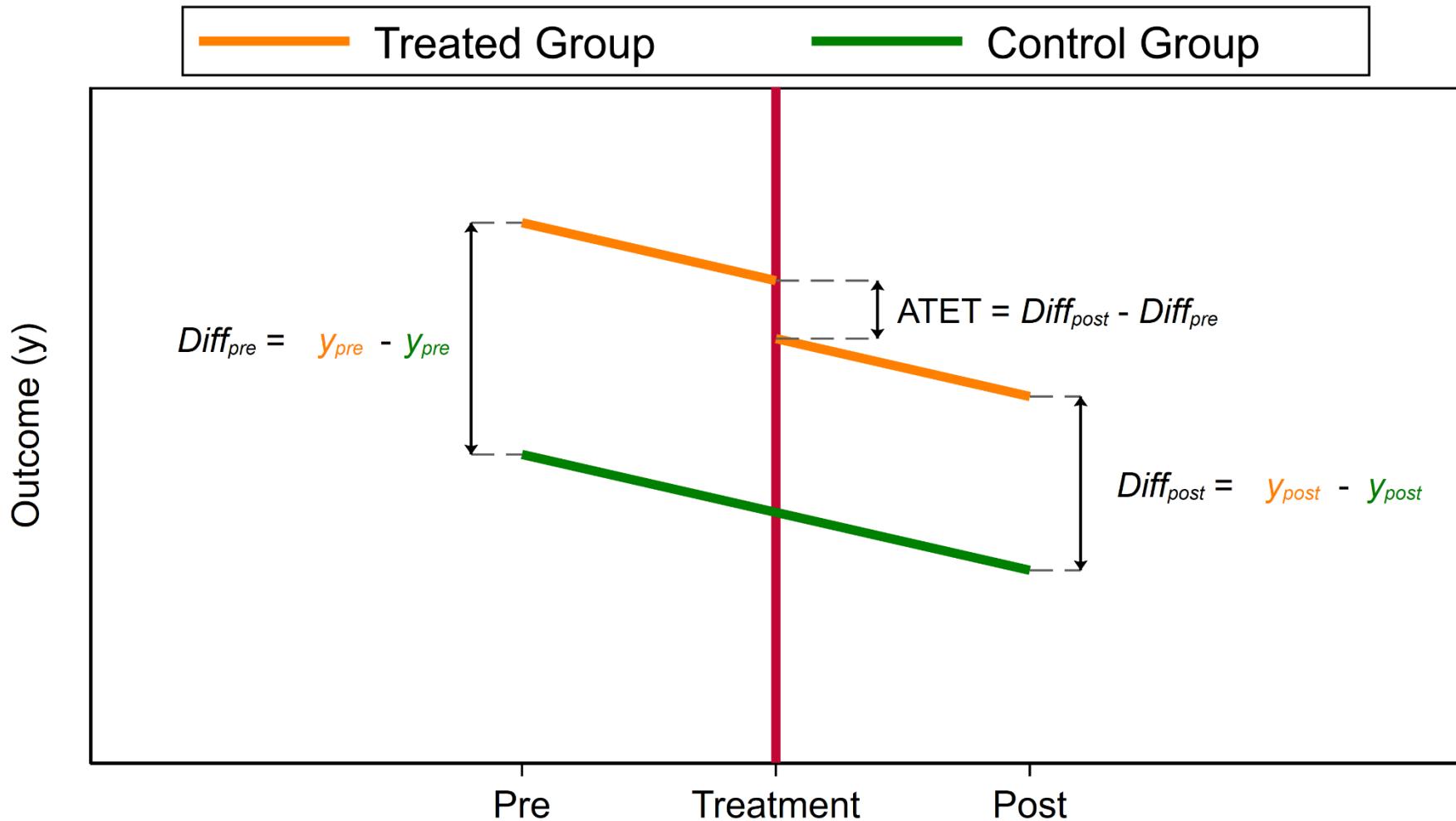
A Conceptual Introduction



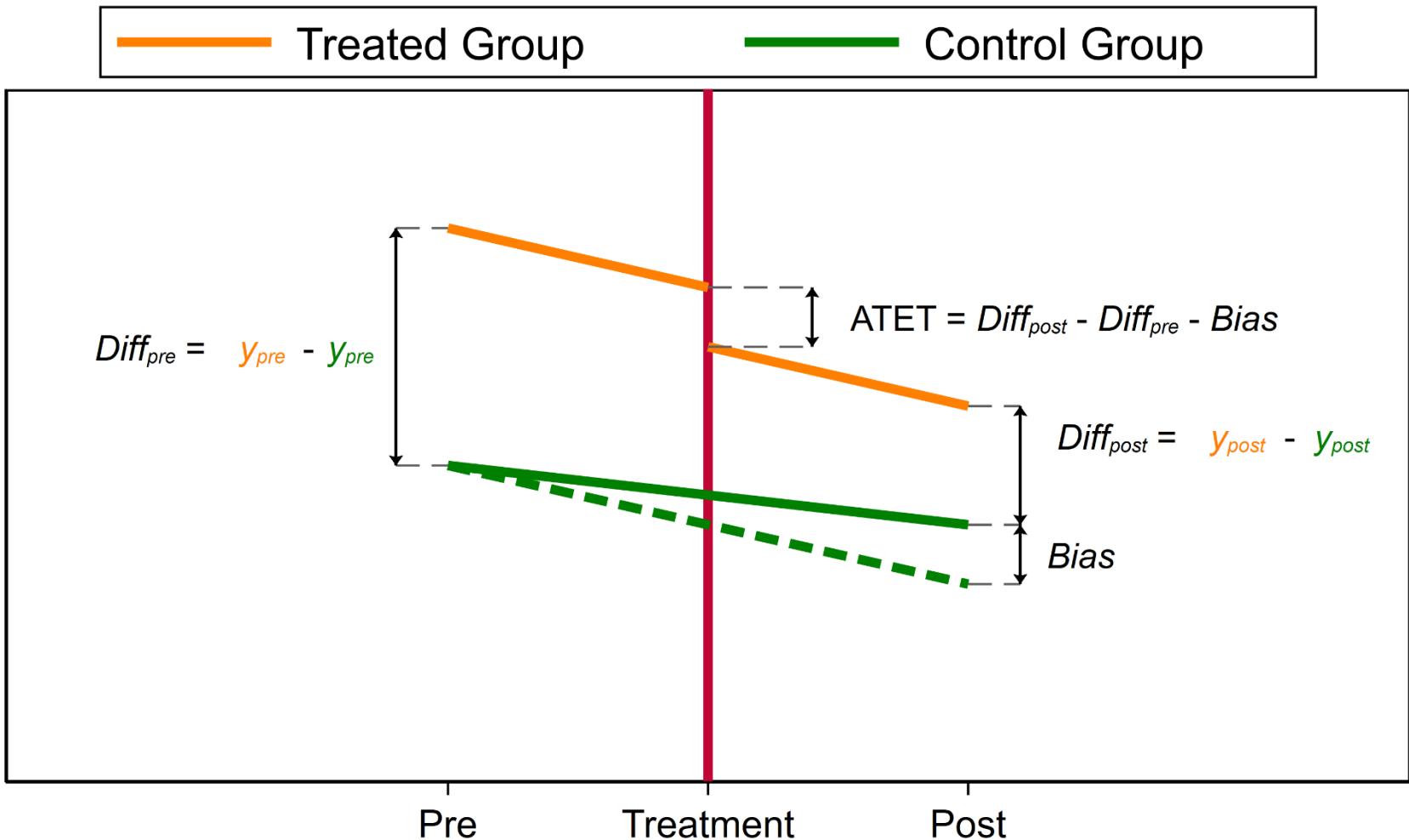
A Conceptual Introduction



A Conceptual Introduction



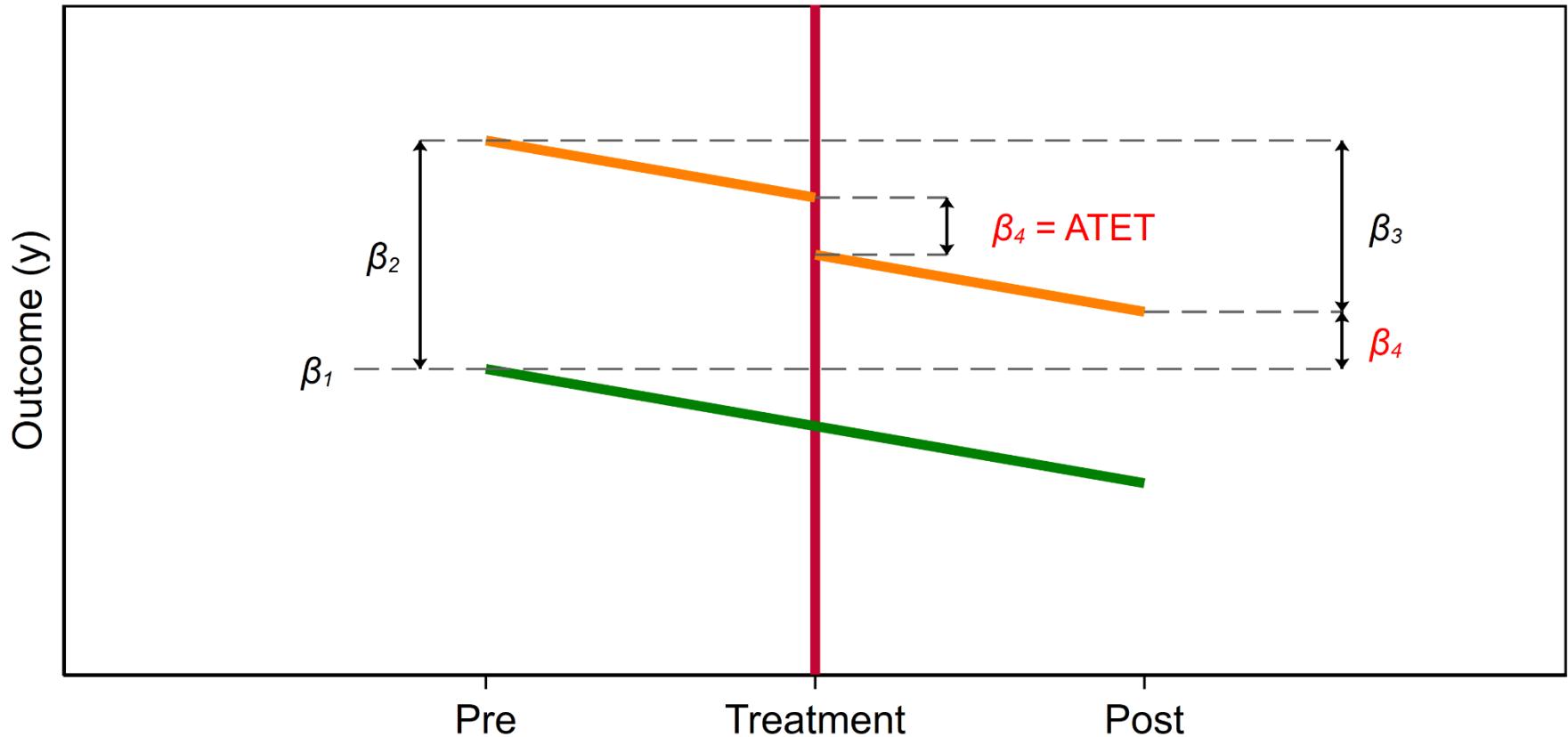
A Conceptual Introduction



A Conceptual Introduction

$$E(y_{it}|\text{Group}, \text{Time}) = \beta_1 + \beta_2 \text{Group} + \beta_3 \text{Time} + \beta_4 \text{Group} \times \text{Time}$$

— Treated Group — Control Group

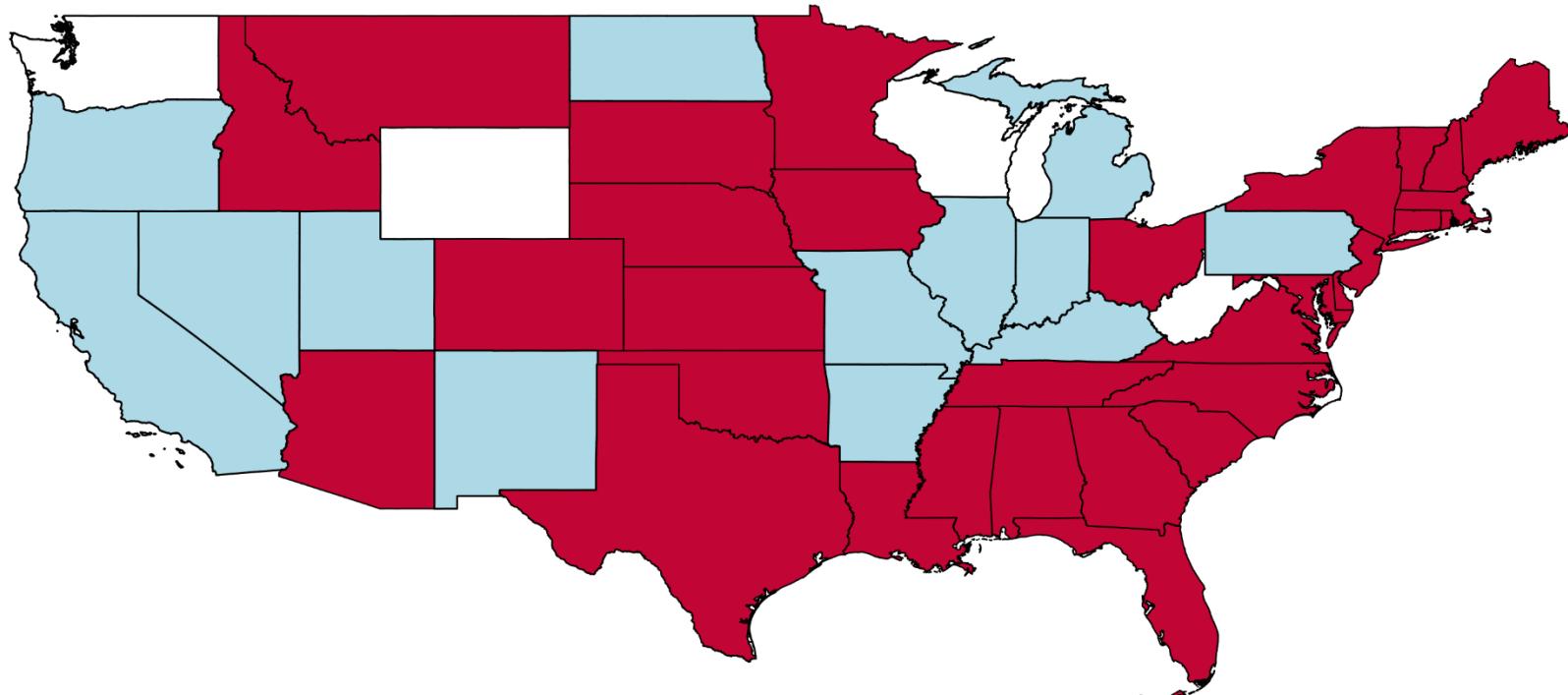


Outline

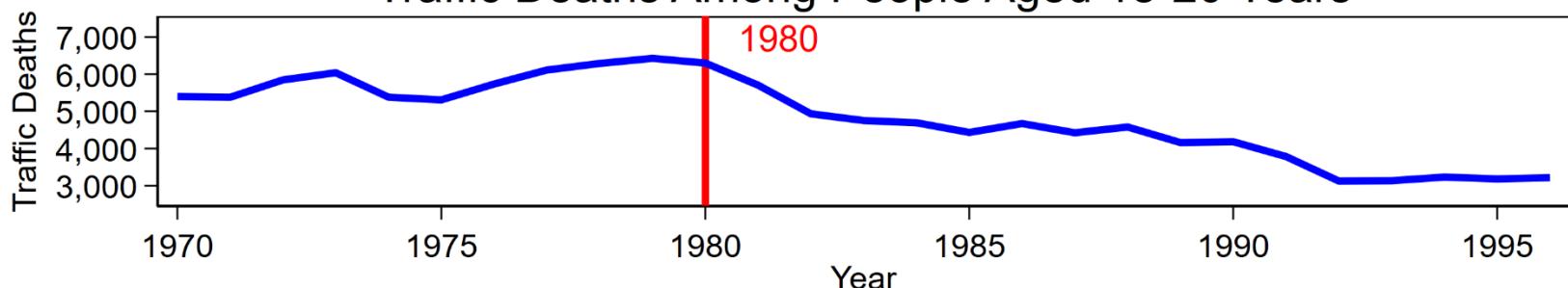
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States With MLDA Less than 21 (red)

Year = 1980

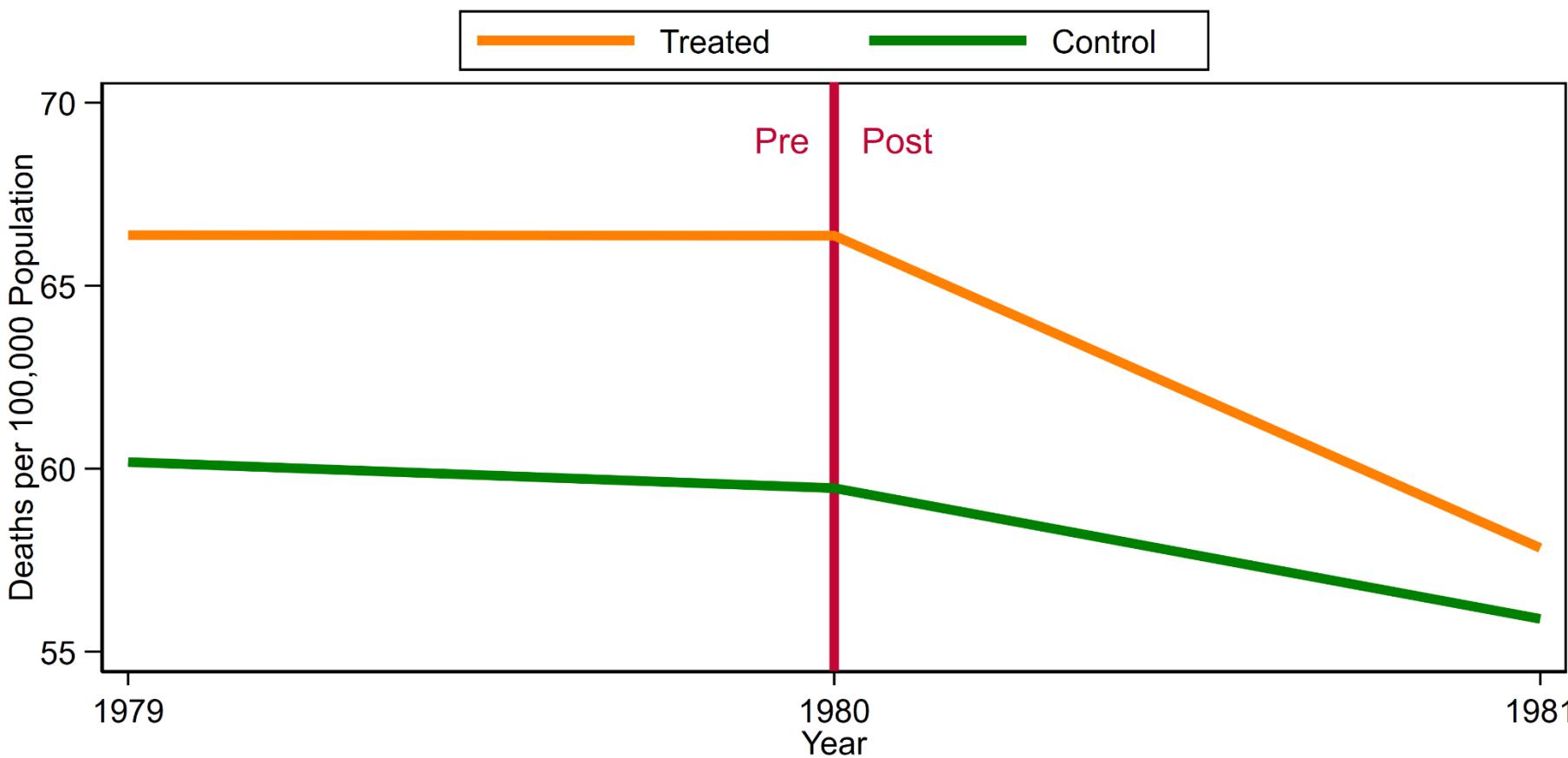


Traffic Deaths Among People Aged 18-20 Years



The Two Period, Two Group Model

Deaths per 100,000 Population Among People 18-20 Years of Age
by MDLA21 Law In Effect Before 1980



The Examples Dataset

```
. use MLDA21, clear  
(Minimum Legal Drinking Age (MLDA) from Angrist and Pischke, 2014 (type 'notes'))  
. describe FIPS state year MLDA21_year MLDA21 mrate_agegrp beertax
```

Variable name	Storage type	Display format	Value label	variable label
FIPS	byte	%10.0g		State ID (Federal Information Processing Standard)
state	str15	%15s		* State
year	int	%9.0g		* Year
MLDA21_year	int	%10.0g		_ Year in which MLDA21 law was enacted
MLDA21	byte	%15.0g	treat83	Pre- or Post- MLDA21 Law
mrate_agegrp	double	%9.2f		* Mortality rate (deaths/100000) in the 18-20 age group
beertax	double	%9.2f		Beer tax

The Examples Dataset

```
. summarize FIPS state year MLDA21_year MLDA21 mrate_agegrp beertax
```

variable	Obs	Mean	Std. dev.	Min	Max
FIPS	1,242	27.13043	14.33362	1	51
state	0				
year	1,242	1983	7.792018	1970	1996
MLDA21_year	1,242	1973.239	21.16342	1933	1988
MLDA21	1,242	.5764895	.4943138	0	1
mrate_agegrp	1,242	52.14309	21.93833	8.231131	189.9421
beertax	1,226	.2367918	.236822	.012747	1.984536

```
. list state MLDA21_year year MLDA21      ///
>      if inlist(FIPS, 17)      ///
>      , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21
Illinois	1980	1970	0
Illinois	1980	1971	0
Illinois	1980	1972	0
Illinois	1980	1973	0
Illinois	1980	1974	0
Illinois	1980	1975	0
Illinois	1980	1976	0
Illinois	1980	1977	0
Illinois	1980	1978	0
Illinois	1980	1979	0
Illinois	1980	1980	1
Illinois	1980	1981	1
Illinois	1980	1982	1
Illinois	1980	1983	1
Illinois	1980	1984	1
Illinois	1980	1985	1
Illinois	1980	1986	1
Illinois	1980	1987	1
Illinois	1980	1988	1
Illinois	1980	1989	1
Illinois	1980	1990	1
Illinois	1980	1991	1
Illinois	1980	1992	1
Illinois	1980	1993	1
Illinois	1980	1994	1
Illinois	1980	1995	1
Illinois	1980	1996	1

```
. // List the data for Iowa
. list state MLDA21_year year MLDA21      ///
>      if inlist(FIPS, 19)      ///
>      , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21
Iowa	1986	1970	0
Iowa	1986	1971	0
Iowa	1986	1972	0
Iowa	1986	1973	0
Iowa	1986	1974	0
Iowa	1986	1975	0
Iowa	1986	1976	0
Iowa	1986	1977	0
Iowa	1986	1978	0
Iowa	1986	1979	0
Iowa	1986	1980	0
Iowa	1986	1981	0
Iowa	1986	1982	0
Iowa	1986	1983	0
Iowa	1986	1984	0
Iowa	1986	1985	0
Iowa	1986	1986	1
Iowa	1986	1987	1
Iowa	1986	1988	1
Iowa	1986	1989	1
Iowa	1986	1990	1
Iowa	1986	1991	1
Iowa	1986	1992	1
Iowa	1986	1993	1
Iowa	1986	1994	1
Iowa	1986	1995	1
Iowa	1986	1996	1

Setting Up The Data

```
. keep if inrange(year, 1979, 1981) } Keep the data for 1979, 1980, and 1981  
(1,104 observations deleted)
```



```
. list state MLDA21_year year MLDA21      ///  
>     if year!=1980 & inlist(FIPS, 17,18,19)      ///  
>     , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21
Illinois	1980	1979	0
Illinois	1980	1981	1
Indiana	1934	1979	1
Indiana	1934	1981	1
Iowa	1986	1979	0
Iowa	1986	1981	0

- } The law went into effect during the time interval
- } The law was in effect during the entire interval
- } The law was never in effect during the entire interval

Setting Up The Data

```
. // Create a variable for time  
. generate time = (year>1980)  
  
. list state MLDA21_year year MLDA21 time      ///  
 >      if year!=1980 & inlist(FIPS, 17,18,19)      ///  
 >      , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21	time
Illinois	1980	1979	0	0
Illinois	1980	1981	1	1
Indiana	1934	1979	1	0
Indiana	1934	1981	1	1
Iowa	1986	1979	0	0
Iowa	1986	1981	0	1

Time equals 0 prior to the intervention in 1980 and time equals 1 after the intervention in 1980

Setting Up The Data

```
// Create a variable for treatment status  
gen treated = .  
replace treated = 0 if MLDA21_year >1980  
replace treated = 0 if MLDA21_year <=1980 & year<1980  
replace treated = 1 if MLDA21_year <=1980  
  
. list state MLDA21_year year MLDA21 time treated ///  
>      if year!=1980 & inlist(FIPS, 17,18,19)      ///  
> , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21	time	treated
Illinois	1980	1979	0	0	1
Illinois	1980	1981	1	1	1
Indiana	1934	1979	1	0	1
Indiana	1934	1981	1	1	1
Iowa	1986	1979	0	0	0
Iowa	1986	1981	0	1	0

Treated equals 1 if the MLDA21 law was in effect in that state during 1980 or sooner. Treated equals 0 if the law went into effect after 1980.

Setting Up The Data

```
. // Create a variable for the interaction of time and treatment  
. gen TimeTreat = time*treated  
  
. list state MLDA21_year year MLDA21 time treated TimeTreat ///  
>     if year!=1980 & inlist(FIPS, 17,18,19)           ///  
> , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21	time	treated	TimeTreat
Illinois	1980	1979	0	0	1	0
Illinois	1980	1981	1	1	1	1
Indiana	1934	1979	1	0	1	0
Indiana	1934	1981	1	1	1	1
Iowa	1986	1979	0	0	0	0
Iowa	1986	1981	0	1	0	0

TimeTreat equals time x treated.

It will be the interaction term when we use regress.

It will be the treatment variable when we use didregress!!!!

DiD Using regress

```
. regress mrate_agegrp time treated TimeTreat, vce(cluster FIPS)
```

Linear regression

Number of obs = 138
F(3, 45) = 5.76
Prob > F = 0.0020
R-squared = 0.0233
Root MSE = 22.879

(Std. err. adjusted for 46 clusters in FIPS)

mrate_agegrp	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
time	-3.92868	2.045822	-1.92	0.061	-8.049176	.1918168
treated	6.549406	6.480692	1.01	0.318	-6.503378	19.60219
TimeTreat	-4.611695	3.32785	-1.39	0.173	-11.31433	2.090939
_cons	59.82451	3.878595	15.42	0.000	52.01262	67.6364

The Average Treatment Effect Among the Treated (ATET) is the coefficient for TimeTreat.

The MLDA21 law caused a decrease of 4.6 people per 100,000 population in the 18-20 year old age group between 1979 and 1981.

DiD Using regress

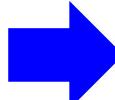
```
. regress mrate_agegrp i.time##i.treated, vce(cluster FIPS)
```

Linear regression

Number of obs = 138
F(3, 45) = 5.76
Prob > F = 0.0020
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1.treated	6.549406	6.480692	1.01	0.318	-6.503378	19.60219
time#treated						
1 1	-4.611695	3.32785	-1.39	0.173	-11.31433	2.090939
_cons	59.82451	3.878595	15.42	0.000	52.01262	67.6364



DiD Using regress

```
. regress mrate_agegrp i.year TimeTreat i.FIPS, vce(cluster FIPS)
```

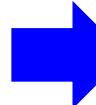
Linear regression

Number of obs = 138
F(2, 45) = .
Prob > F = .
R-squared = 0.8714
Root MSE = 10.187

(std. err. adjusted for 46 clusters in FIPS)

mrate_agegrp	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]	
year						
1980	-.5155614	2.882514	-0.18	0.859	-6.321242	5.29012
1981	-4.18646	3.313667	-1.26	0.213	-10.86053	2.487608
TimeTreat	-4.611695	4.083392	-1.13	0.265	-12.83607	3.612679
FIPS						
2	10.37779	2.88e-13	3.6e+13	0.000	10.37779	10.37779
4	3.268883	2.88e-13	1.1e+13	0.000	3.268883	3.268883
5	5.474611	1.361131	4.02	0.000	2.733153	8.216069

(FIPS output truncated)



DiD Using regress

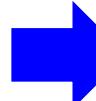
```
. regress mrate_agegrp i.year TimeTreat i.FIPS, vce(cluster FIPS)
```

Linear regression

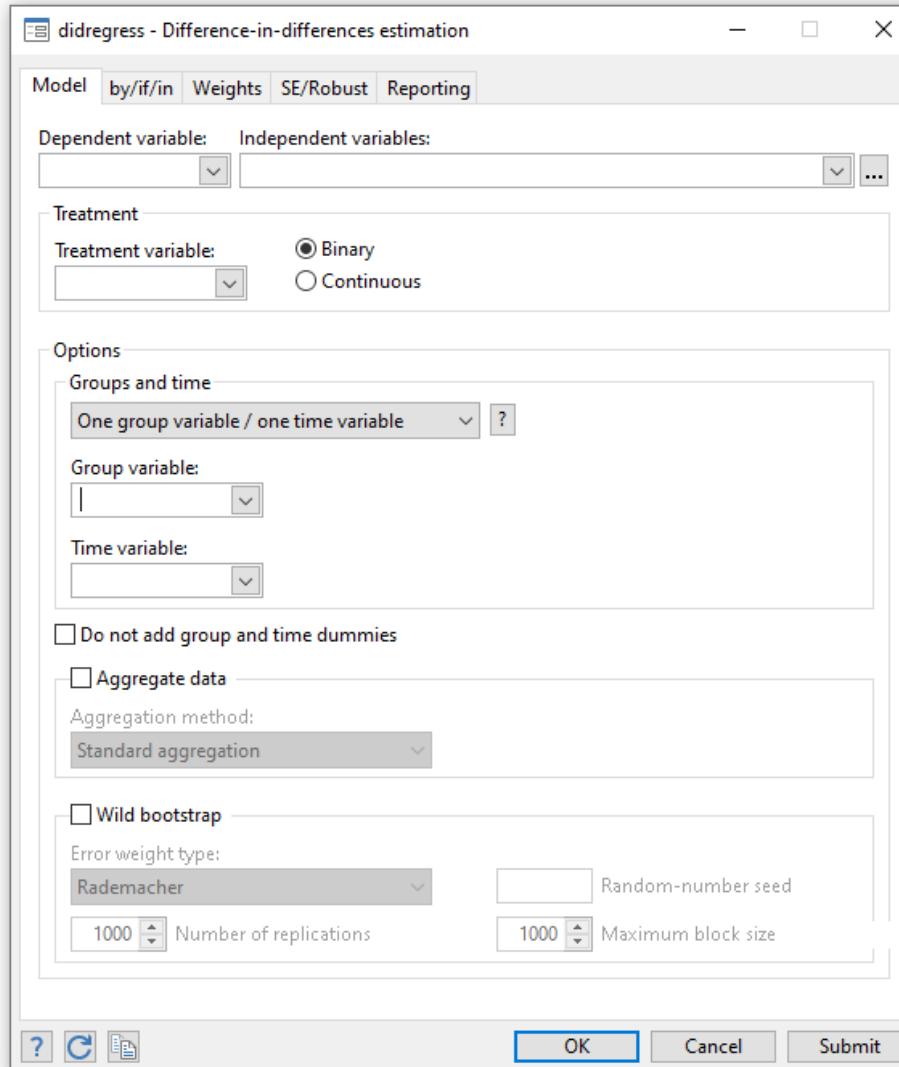
Number of obs = 138
F(2, 45) = .
Prob > F = .
R-squared = 0.8714
Root MSE = 10.187

(std. err. adjusted for 46 clusters in FIPS)

mrate_agegrp	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]
year					
1980	-.5155614	2.882514	-0.18	0.859	-6.321242 5.29012
1981	-4.18646	3.313667	-1.26	0.213	-10.86053 2.487608
TimeTreat	-4.611695	4.083392	-1.13	0.265	-12.83607 3.612679
FIPS					
2	10.37779	2.88e-13	3.6e+13	0.000	10.37779 10.37779
4	3.268883	2.88e-13	1.1e+13	0.000	3.268883 3.268883
5	5.474611	1.361131	4.02	0.000	2.733153 8.216069



DiD Using regress



DiD Using didregress

```
didregress (ovar omvarlist)           ///
            (tvar[, continuous]),      ///
            group(groupvars)          ///
            [time(timevar) options]
```

```
didregress (mrate_agegrp)           ///
            (TimeTreat),             ///
            group(FIPS)              ///
            time(year)
```

TimeTreat = Time x Treat !!!!

DiD Using didregress

```
. didregress (mrate_agegrp)(TimeTreat), group(FIPS) time(year)
```

Number of groups and treatment time

Time variable: year

Control: TimeTreat = 0

Treatment: TimeTreat = 1

	Control	Treatment
Group		
FIPS	33	13
Time		
Minimum	1979	1981
Maximum	1979	1981

Difference-in-differences regression

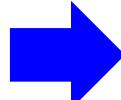
Number of obs = 138

Data type: Repeated cross-sectional

(Std. err. adjusted for 46 clusters in FIPS)

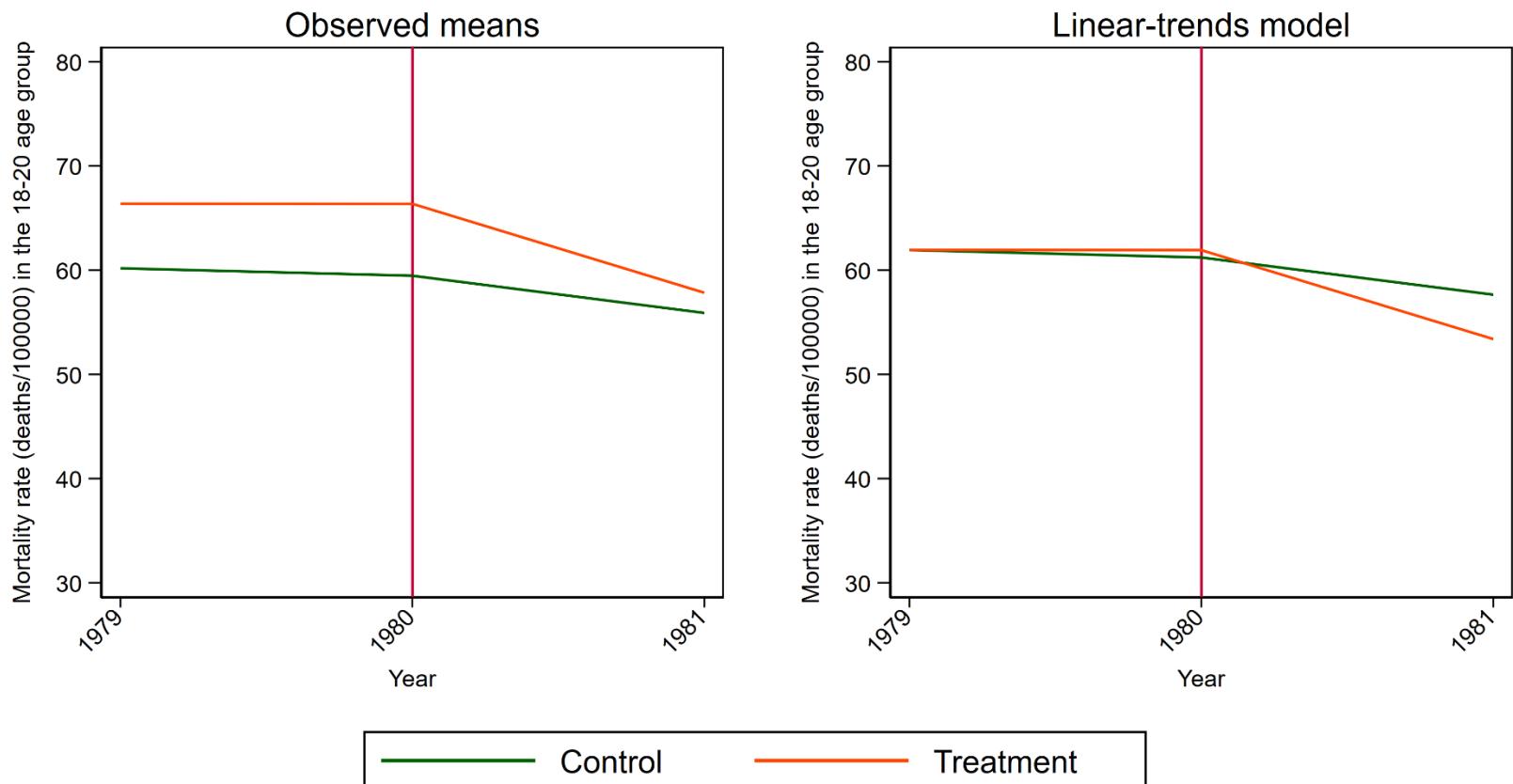
mrate_agegrp	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]
ATET TimeTreat (1 vs 0)	-4.611695	4.083392	-1.13	0.265	-12.83607 3.612679

Note: ATET estimate adjusted for group effects and time effects.



Checking the Parallel Trends Assumption

Graphical diagnostics for parallel trends



estat trendplots

Checking the Parallel Trends Assumption

```
. estat granger
```

Granger causality test

H0: No effect in anticipation of treatment

F(1, 45) = 0.02

Prob > F = 0.8871

didregress With Covariates

```
. didregress (mrate_agegrp beertax) (TimeTreat), group(FIPS) time(year)
```

Number of groups and treatment time

Time variable: year
 Control: TimeTreat = 0
 Treatment: TimeTreat = 1

	Control	Treatment
Group		
FIPS	32	13
Time		
Minimum	1979	1981
Maximum	1979	1981

Difference-in-differences regression
 Data type: Repeated cross-sectional

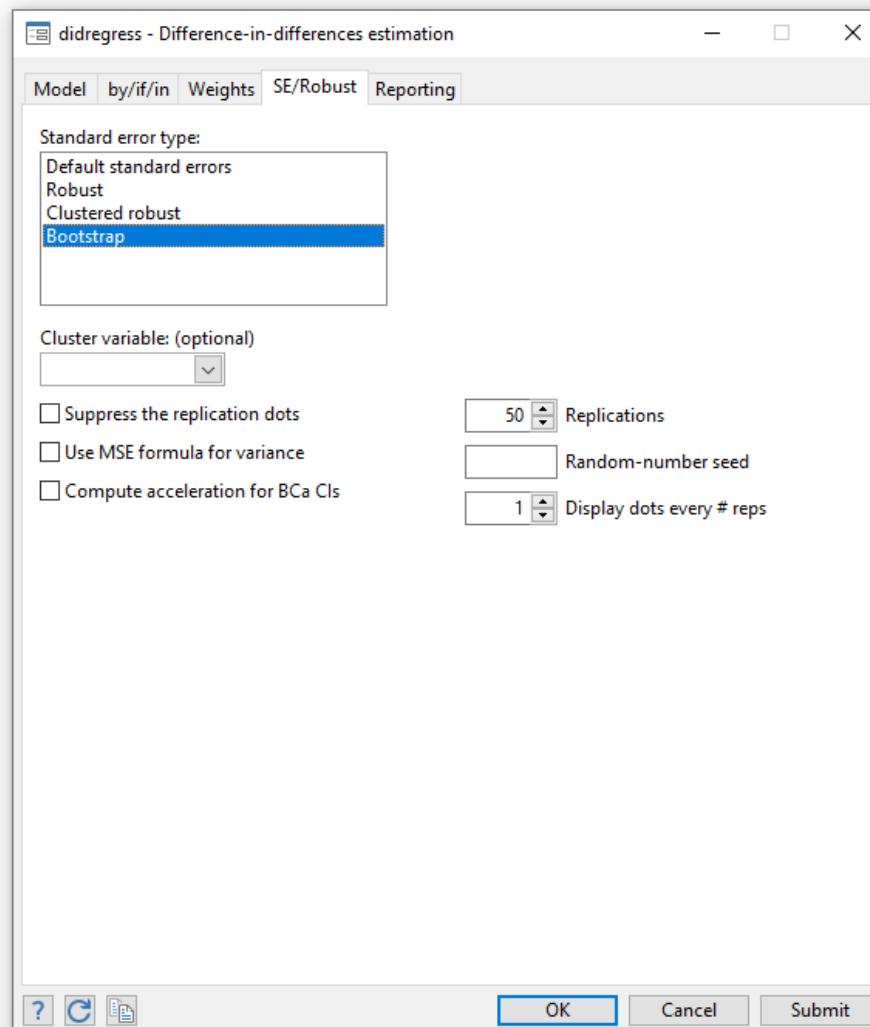
Number of obs = 135

(Std. err. adjusted for 45 clusters in FIPS)

mrate_agegrp	Coefficient	Robust std. err.	t	P> t	[95% conf. interval]
ATET TimeTreat (1 vs 0)	-5.140093	4.195906	-1.23	0.227	-13.59639 3.316199

Note: ATET estimate adjusted for covariates, group effects, and time effects.

Estimation of Standard Errors



Estimation of Standard Errors

Must account for the serial correlation of the outcome.

For a large number of groups

By default, `didregress` uses cluster–robust standard errors and uses critical values of a t distribution with $G - 1$ degrees of freedom, where G is the number of groups.

For more information: <https://www.stata.com/manuals/tedidintro.pdf>

Estimation of Standard Errors

For a small number of groups

`wildbootstrap[(wildopts)]` computes confidence intervals and *p*-values with the wild bootstrap. The wild bootstrap is constructed imposing the null hypothesis that the ATET is 0; that is, it is a restricted wild bootstrap. Confidence intervals are computed separately from the *p*-values. The bounds of the confidence interval are computed using a bisection optimization algorithm described in Methods and formulas in [TE] `didregress`. *wildopts* are `errorweight(edtype)`, `reps(#)`, `rseed(#)`, and `blocksize(#)`.

`errorweight(edtype)` defines the error weight used to draw residuals from the wild bootstrap. *edtype* is one of `rademacher` (the default), `mammen`, `webb`, `normal`, or `gamma`.

`rademacher` multiplies the residuals at each bootstrap replication with a randomly generated variable that takes the value of 1 with probability 0.5 and the value of -1 with probability 0.5. `errorweight(rademacher)` is the default.

`mammen` multiplies the residuals at each bootstrap replication with a randomly generated variable that takes the value of 1 - *phi* with probability *phi*/ $\sqrt{5}$ and *phi* otherwise, where *phi* = $(1+\sqrt{5})/2$.

`webb` multiplies the residuals at each bootstrap replication with a randomly generated variable that takes the values $-\sqrt{3}/2$, $-\sqrt{2}/2$, $-\sqrt{1}/2$, $\sqrt{1}/2$, $\sqrt{2}/2$, and $\sqrt{3}/2$, each with probability 1/6.

`normal` multiplies the residuals at each bootstrap replication with a randomly generated normal distribution variable with the first four moments given by 0, 1, 0, and 3.

`gamma` multiplies the residuals at each bootstrap replication with a randomly generated gamma distribution variable with shape parameter 4 and scale parameter 1/2.

`reps(#)` performs # wild bootstrap replications. The default is `reps(1000)`.

`rseed(#)` sets the random-number seed to #.

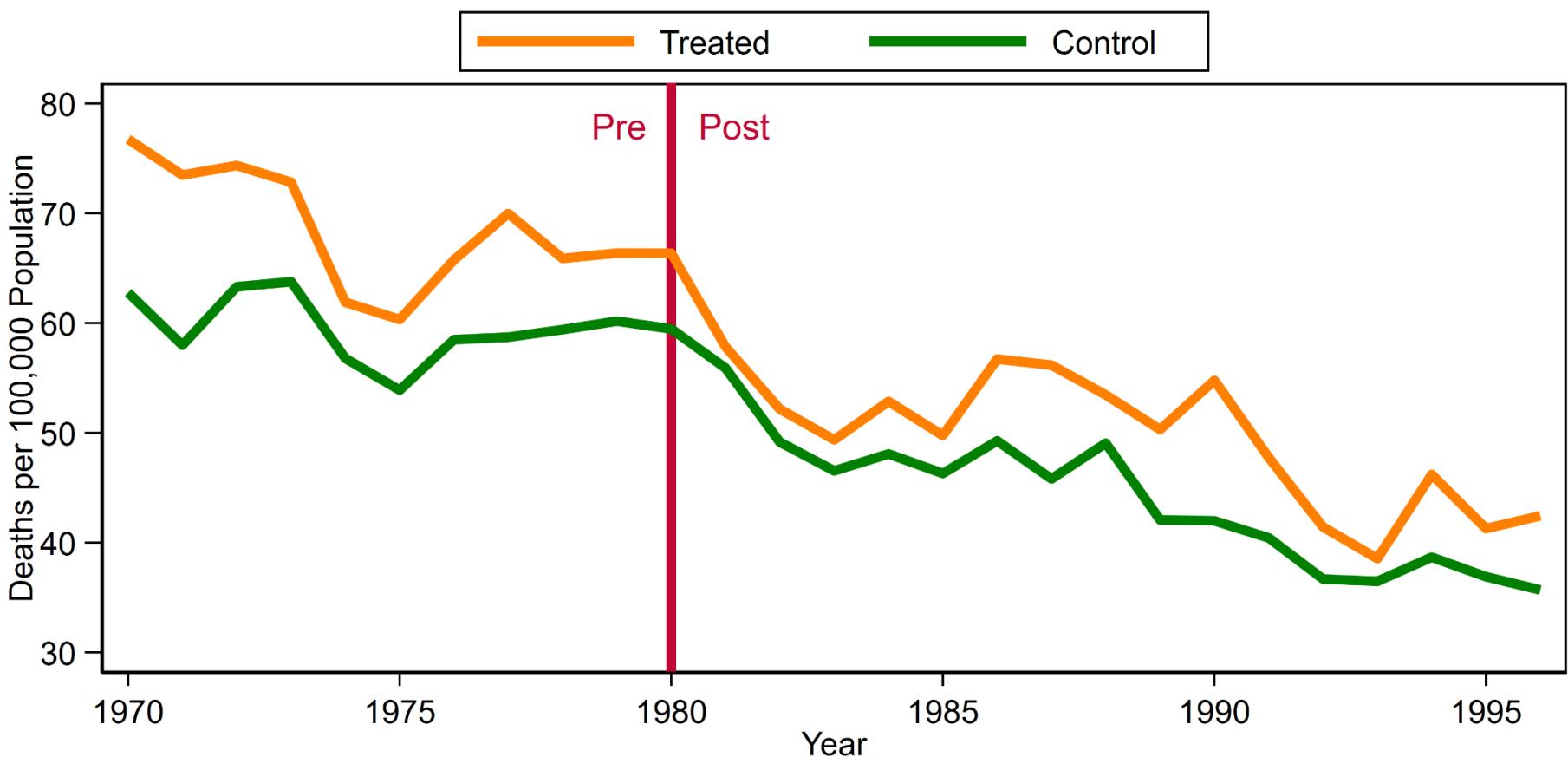
`blocksize(#)` specifies that the wild bootstrap be performed in blocks, with # replications per block. The wild bootstrap computation requires a matrix with dimensions (# groups) x (# replications). If this is too large, you can reduce the matrix to (# groups) x (# block size) and loop (# replications)/(# block size) times. When the same random seed is set, using a different block size does not change the numerical results; it only modifies the computation method. The block size must be less than or equal to the number of bootstrap replications.

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DiD For Repeated Cross-Sectional Data

Deaths per 100,000 Population Among People 18-20 Years of Age
by MDLA21 Law In Effect Before 1980



DiD For Repeated Cross-Sectional Data

```
use MLDA21, clear
```

```
// Create a variable for time  
generate time = (year>1980)
```

```
// Create a variable for treatment status  
gen treated = .  
replace treated = 0 if MLDA21_year >1980  
replace treated = 0 if MLDA21_year <=1980 & year<1980  
replace treated = 1 if MLDA21_year <=1980
```

```
// Create a variable for the interaction of time and treatment  
gen TimeTreat = time*treated
```

```
. list state MLDA21_year year MLDA21 time treated TimeTreat ///
>     if year!=1980 & inlist(FIPS, 17) ///
>     , sepby(state) nolabel noobs abbrev(18)
```

state	MLDA21_year	year	MLDA21	time	treated	TimeTreat
Illinois	1980	1970	0	0	1	0
Illinois	1980	1971	0	0	1	0
Illinois	1980	1972	0	0	1	0
Illinois	1980	1973	0	0	1	0
Illinois	1980	1974	0	0	1	0
Illinois	1980	1975	0	0	1	0
Illinois	1980	1976	0	0	1	0
Illinois	1980	1977	0	0	1	0
Illinois	1980	1978	0	0	1	0
Illinois	1980	1979	0	0	1	0
Illinois	1980	1981	1	1	1	1
Illinois	1980	1982	1	1	1	1
Illinois	1980	1983	1	1	1	1
Illinois	1980	1984	1	1	1	1
Illinois	1980	1985	1	1	1	1
Illinois	1980	1986	1	1	1	1
Illinois	1980	1987	1	1	1	1
Illinois	1980	1988	1	1	1	1
Illinois	1980	1989	1	1	1	1
Illinois	1980	1990	1	1	1	1
Illinois	1980	1991	1	1	1	1
Illinois	1980	1992	1	1	1	1
Illinois	1980	1993	1	1	1	1
Illinois	1980	1994	1	1	1	1
Illinois	1980	1995	1	1	1	1
Illinois	1980	1996	1	1	1	1

DiD For Repeated Cross-Sectional Data

```
. didregress (mrate_agegrp)(TimeTreat), group(FIPS) time(year)
```

Number of groups and treatment time

Time variable: year

Control: TimeTreat = 0

Treatment: TimeTreat = 1

	Control	Treatment
Group		
FIPS	33	13
Time		
Minimum	1970	1981
Maximum	1970	1981

Difference-in-differences regression

Number of obs = 1,242

Data type: Repeated cross-sectional

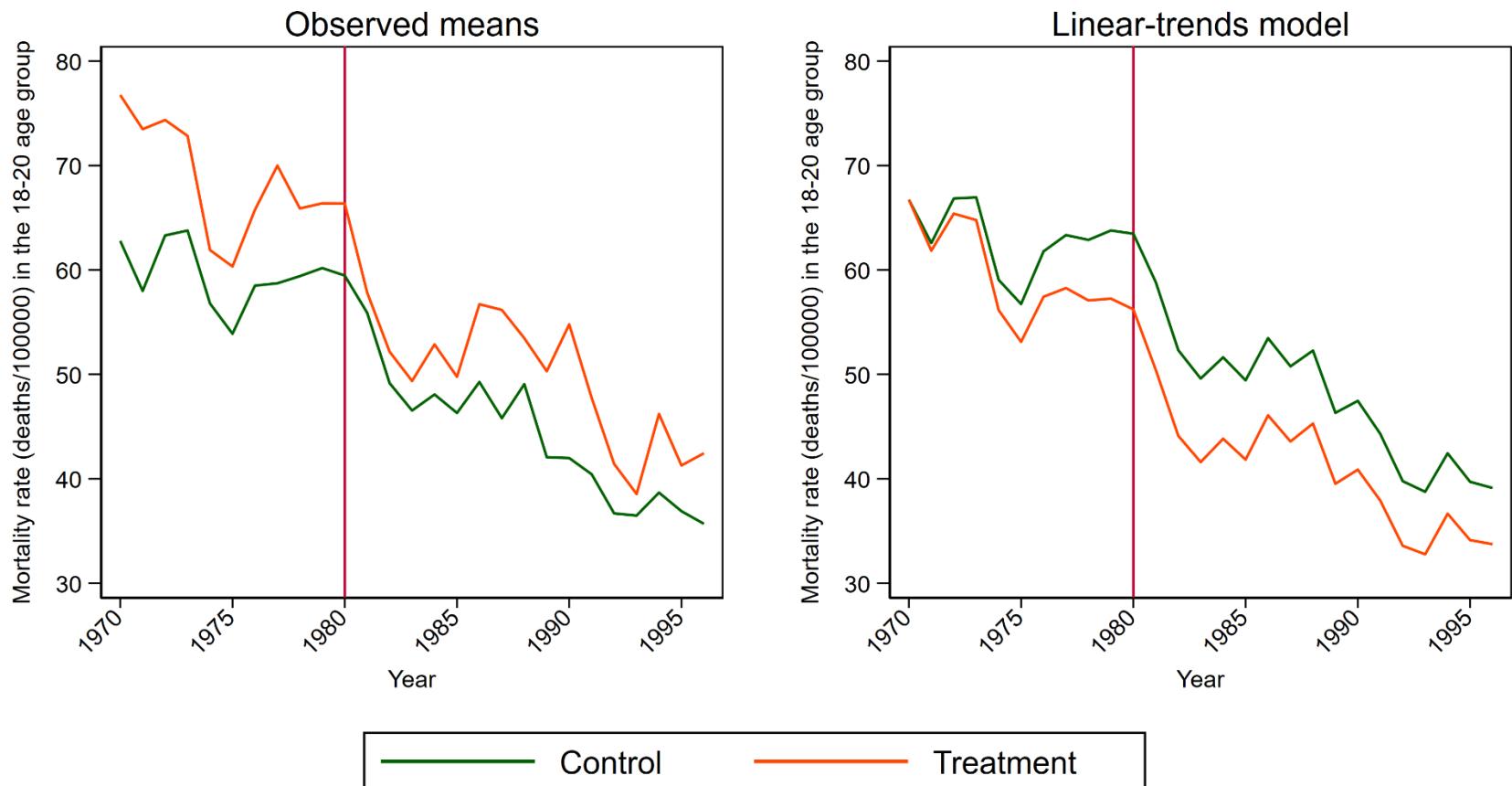
(Std. err. adjusted for 46 clusters in FIPS)

mrate_agegrp	Robust				
	Coefficient	std. err.	t	P> t	[95% conf. interval]
ATET TimeTreat (1 vs 0)	-3.268054	4.212866	-0.78	0.442	-11.7532 5.217093

Note: ATET estimate adjusted for group effects and time effects.

DiD For Repeated Cross-Sectional Data

Graphical diagnostics for parallel trends



estat trendplots

DiD For Repeated Cross-Sectional Data

. estat granger

Granger causality test

H0: No effect in anticipation of treatment

F(10, 45) = 0.95

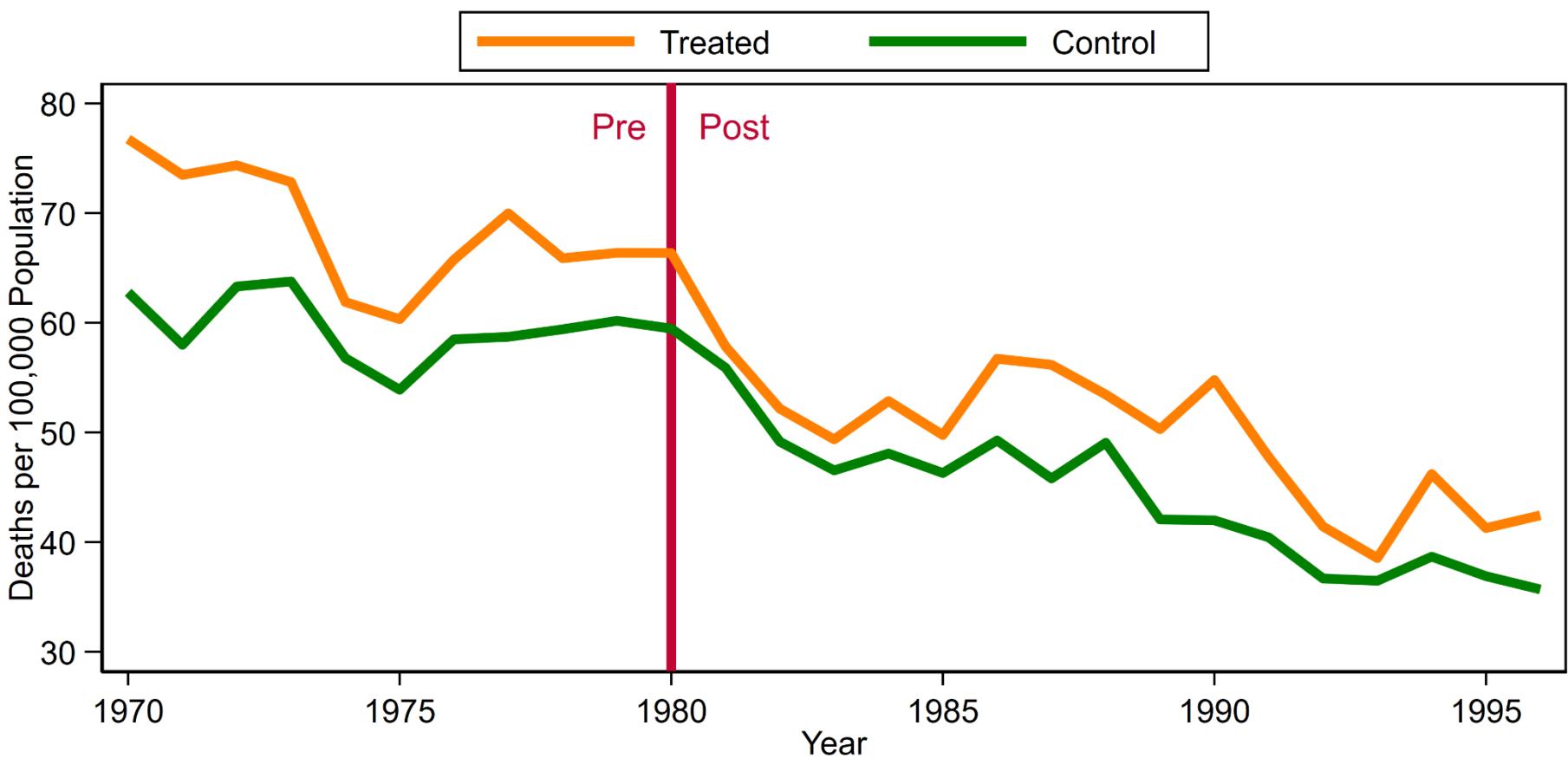
Prob > F = 0.4952

Outline

- ✓ • The Question
- ✓ • An Intuitive Introduction
- ✓ • Two Period, Two Groups Model
- ✓ • Repeated Cross-Sectional Panel Data
 - Longitudinal Panel Data
 - More information

DiD For Longitudinal Panel Data

Deaths per 100,000 Population Among People 18-20 Years of Age
by MDLA21 Law In Effect Before 1980



DiD For Longitudinal/Panel Data

```
. xtset FIPS year, yearly
```

Panel variable: FIPS (strongly balanced)

Time variable: year, 1970 to 1996
Delta: 1 year

DiD For Longitudinal/Panel Data

```
. xtdidregress (mrate_agegrp)(TimeTreat), group(FIPS) time(year)
```

Number of groups and treatment time

Time variable: year

Control: TimeTreat = 0

Treatment: TimeTreat = 1

	Control	Treatment
Group		
FIPS	33	13
Time		
Minimum	1970	1981
Maximum	1970	1981

Difference-in-differences regression

Number of obs = 1,242

Data type: Longitudinal

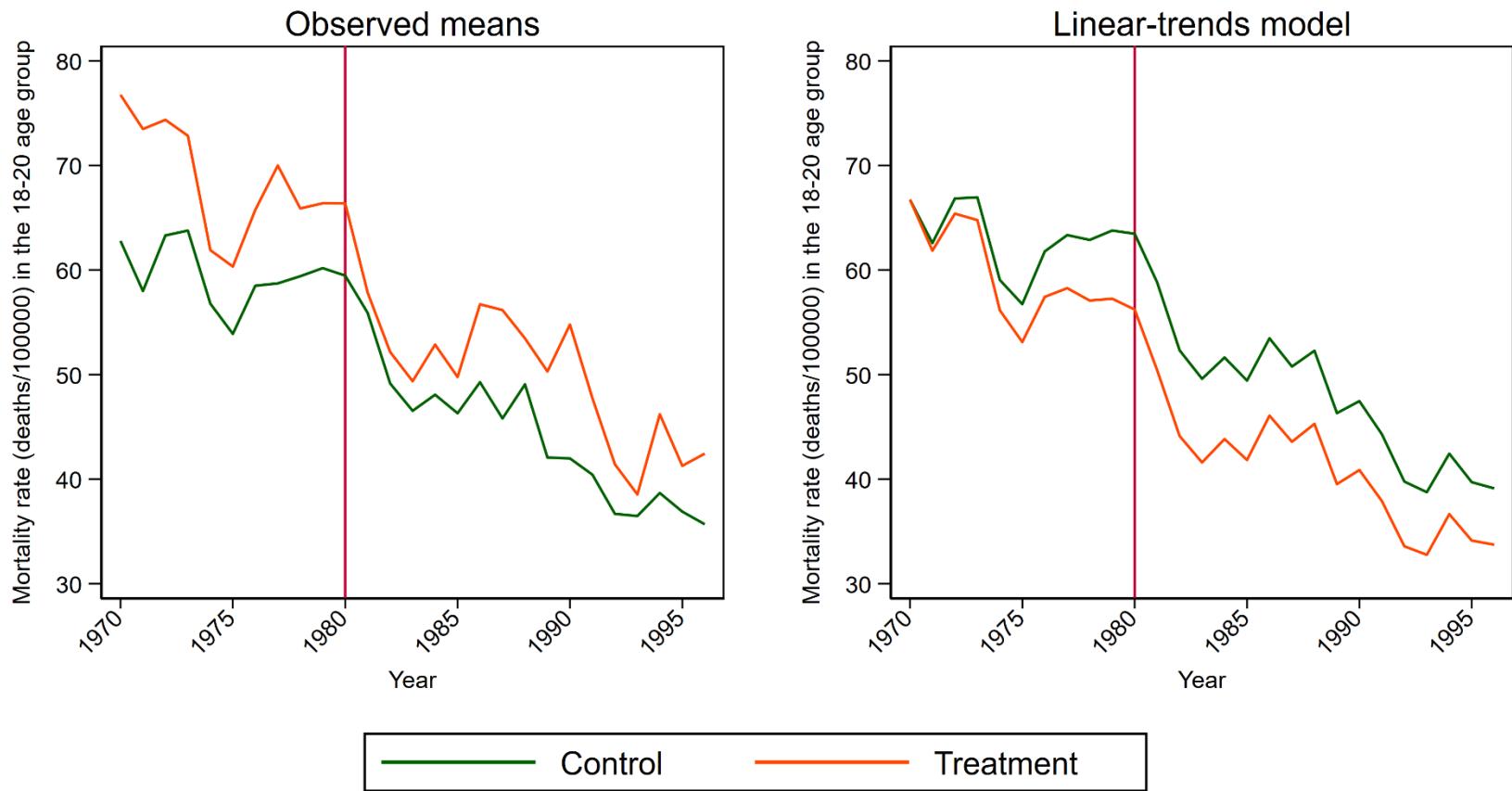
(Std. err. adjusted for 46 clusters in FIPS)

mrate_agegrp	Robust					
	Coefficient	std. err.	t	P> t	[95% conf. interval]	
ATET TimeTreat (1 vs 0)	-3.268054	4.134048	-0.79	0.433	-11.59445	5.058346

Note: ATET estimate adjusted for panel effects and time effects.

DiD For Longitudinal/Panel Data

Graphical diagnostics for parallel trends



estat trendplots

DiD For Longitudinal/Panel Data

. estat granger

Granger causality test

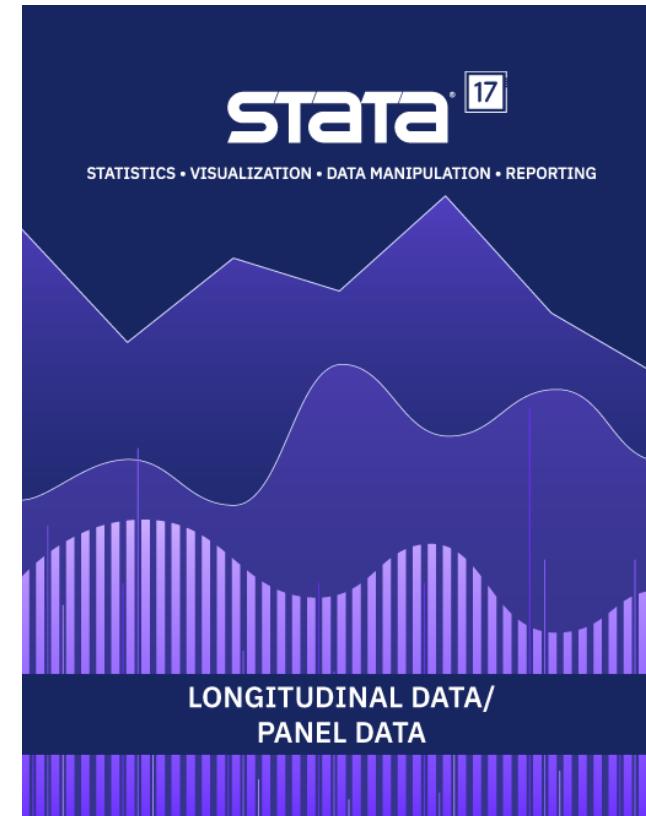
H0: No effect in anticipation of treatment

F(10, 45) = 0.99
Prob > F = 0.4650

Outline

- ✓ • The Question
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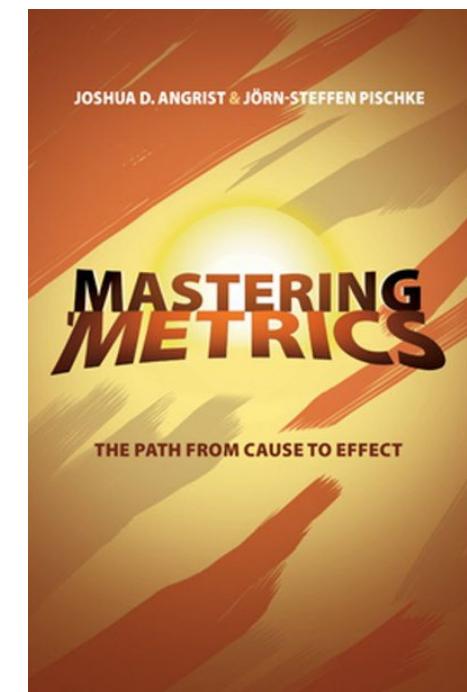
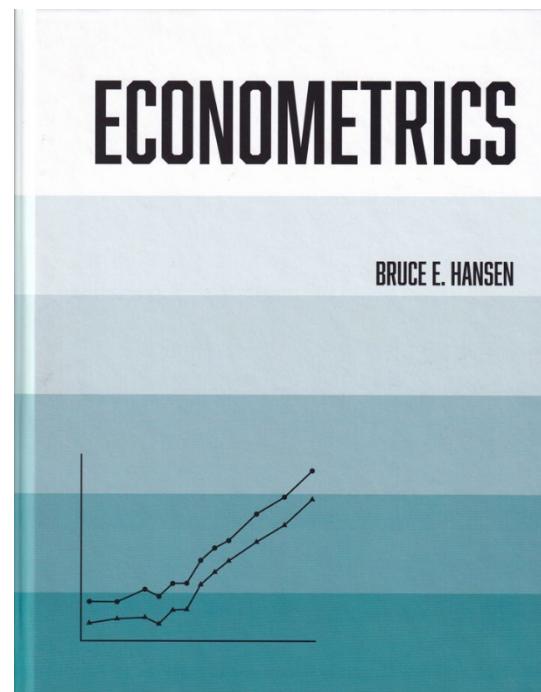
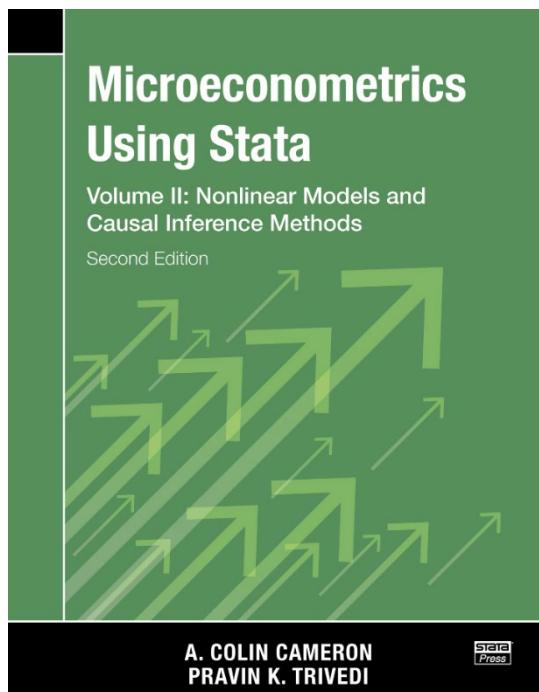
Read The Fine Manuals!



didregress <https://www.stata.com/manuals/te.pdf>

xtdidregress <https://www.stata.com/manuals/xt.pdf>

More Information



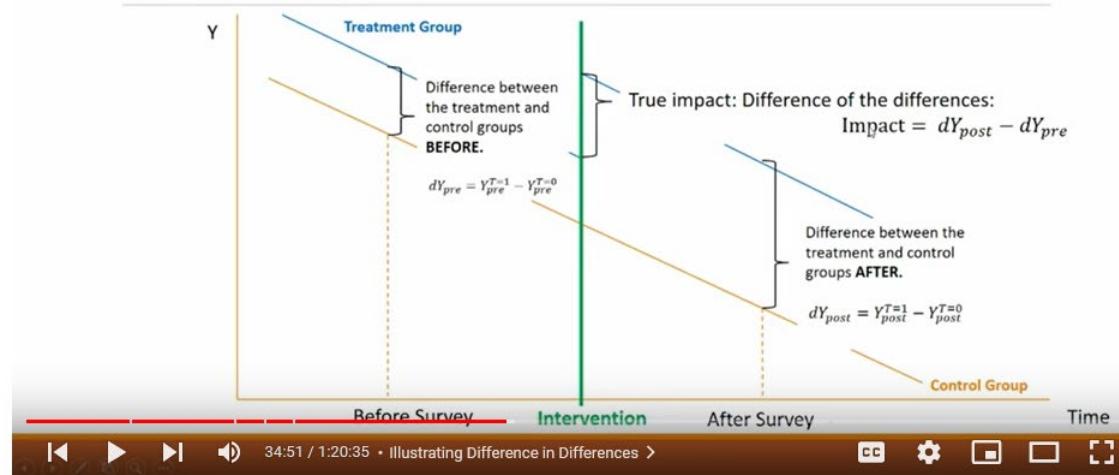
More Information



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Illustrating Difference in Differences



Lecture 14 Difference in Differences

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<https://www.youtube.com/watch?v=rrgxR97QBpc>

More Information

[US National Highway Traffic Safety Administration Information](#)

[Miron and Tetelbaum, 2009](#)

Thank you!

Questions?

You can download the slides, datasets, and do-files here:

<https://tinyurl.com/StataDID>

You can contact me anytime at chuber@stata.com