

Support Violations for Difference-in-Differences

Adam N. Glynn ¹ Nahomi Ichino ¹ Juraj Medzihorsky ²

¹Emory University

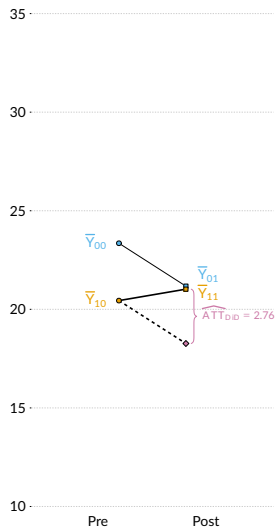
²Durham University

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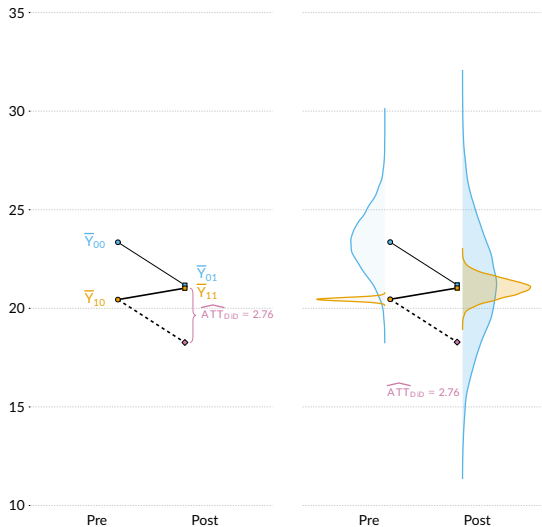
Support Issues for 2-period DiD

1. Adding distributional information on the outcomes to PT plots can reveal situations where quantile-based CiC (Athey and Imbens 2006) will be preferred to standard DiD.
2. Plots of imputed PO distributions to assess the robustness of ATT findings.
3. The ATT can be bounded by a support-adjusted estimate.
4. Because time-invariant and group-invariant confounding are not equivalent assumptions on the quantile scale, simultaneous presentation of CiC and reverse CiC, in addition to support-adjusted estimates for both, allow for a more robust check on DiD when one of these assumptions is not preferred to the other.

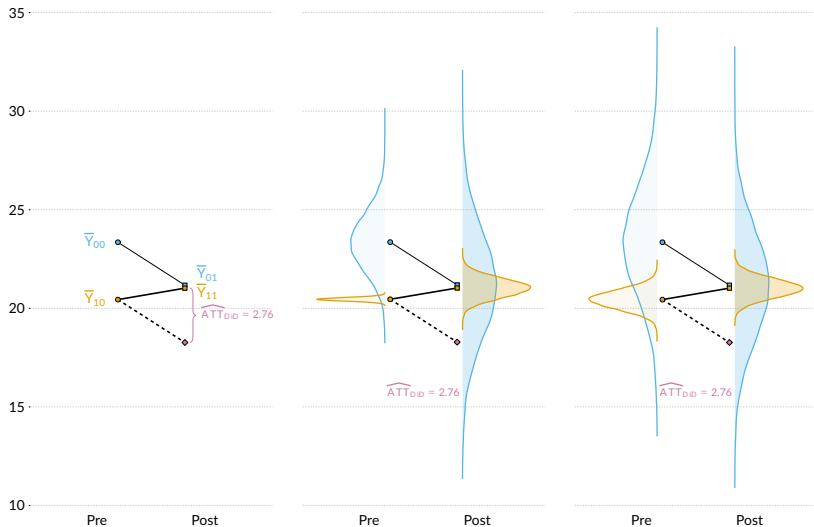
Mean Based Parallel Trends Plot Shows Large Effect



Some Consistent Distributions also Imply Large Effect



Some Consistent Distributions ImPLY Small Effect



Minimum Wage Notation and Decomposed Estimators

As our example, we use the Card and Krueger (1994) analysis of full-time equivalent (FTE) employment in fast food stores in NJ and PA before and after the 1992 NJ minimum wage increase.

Y_{q11} : FTE employment in restaurant q in NJ after the min wage increase.

Y_{r10} : FTE employment in restaurant r in NJ before the min wage increase.

Y_{s01} : FTE employment in restaurant s in PA after the min wage increase.

Y_{t00} : FTE employment in restaurant t in PA before the min wage increase.

Under the linear DiD estimator:

$$\widehat{ATT}_{DiD} = \frac{1}{n_{11}} \sum_{q=1}^{n_{11}} Y_{q11} - \frac{1}{n_{10}} \sum_{r=1}^{n_{10}} \left\{ \underbrace{Y_{r10} + \left[\frac{1}{n_{01}} \sum_{s=1}^{n_{01}} Y_{s01} - \frac{1}{n_{00}} \sum_{t=1}^{n_{00}} Y_{t00} \right]}_{\text{Imputed Potential Outcome}} \right\}$$

Minimum Wage Notation and Decomposed Estimators

Assuming time-invariant confounding on the quantile scale (Sofer et al. 2016):

$$F_{Y_{01}} \left(F_{Y_{11}(0)}^{-1}(\nu) \right) = F_{Y_{00}} \left(F_{Y_{10}}^{-1}(\nu) \right), \nu \in [0, 1]$$

where a (group) = 1 for NJ, = 0 for PA, and t (time) = 1 for after, = 0 for before the minimum wage increase, Athey and Imbens (2006) propose:

$$\widehat{ATT}_{CiC} = \frac{1}{n_{11}} \sum_{q=1}^{n_{11}} Y_{q11} - \frac{1}{n_{10}} \sum_{r=1}^{n_{10}} \left\{ \underbrace{\widehat{F}_{Y_{01}}^{-1} \left(\widehat{F}_{Y_{00}}(Y_{r10}) \right)}_{\text{Imputed Potential Outcome}} \right\}$$

where

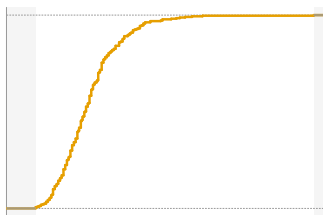
n_{at} : number of restaurants in group a at time t

$\widehat{F}_{Y_{at}}^{-1}$: inverse empirical CDF for employment in group a at time t .

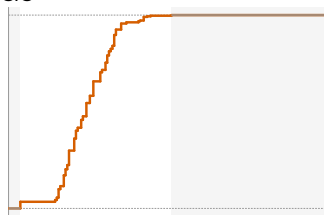
$\widehat{F}_{Y_{at}}$: empirical CDF for employment in group a at time t .

Loop Plot: Changes-in-Changes

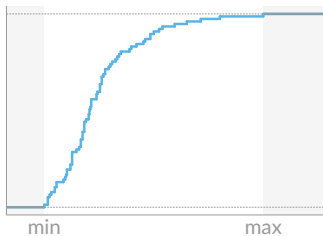
Y_{10}



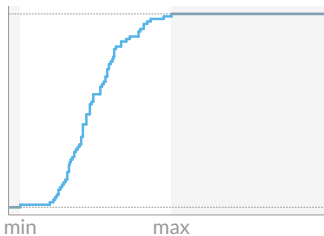
$\widehat{Y_{11}(0)}_{\text{CiC}}$



Y_{00}

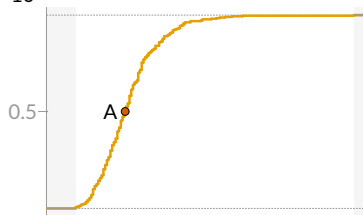


Y_{01}

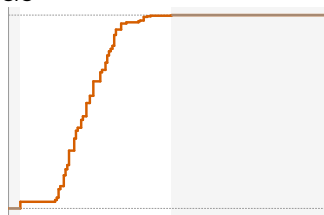


Loop Plot: Changes-in-Changes

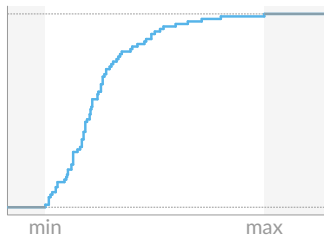
Y_{10}



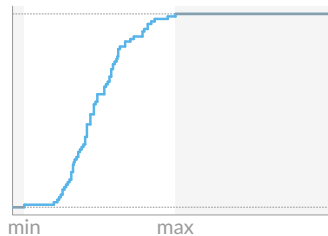
$\widehat{Y_{11}(0)}_{\text{CiC}}$



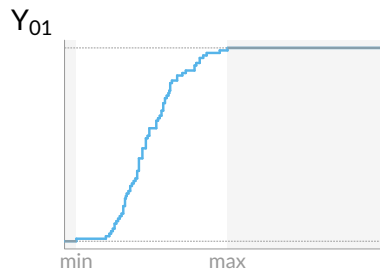
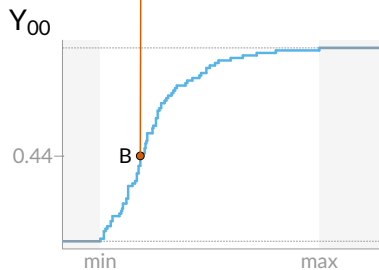
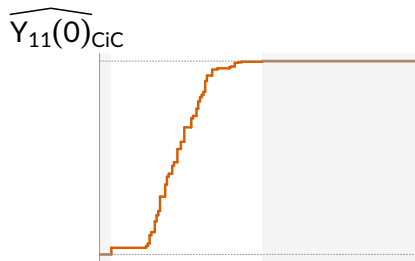
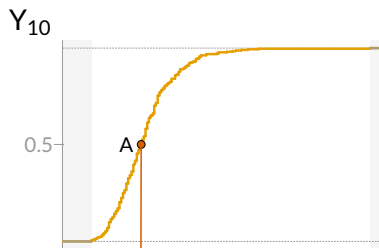
Y_{00}



Y_{01}

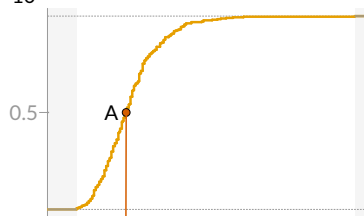


Loop Plot: Changes-in-Changes

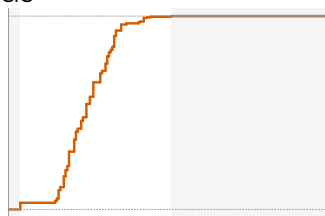


Loop Plot: Changes-in-Changes

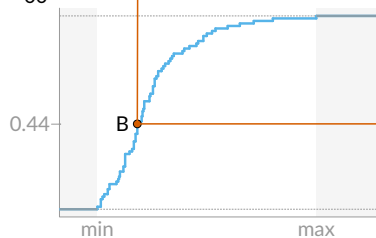
Y_{10}



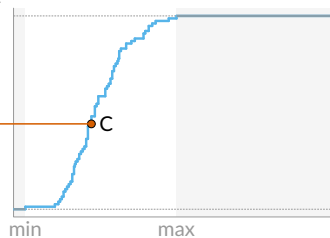
$\widehat{Y_{11}(0)}_{CiC}$



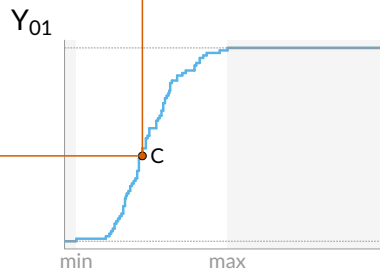
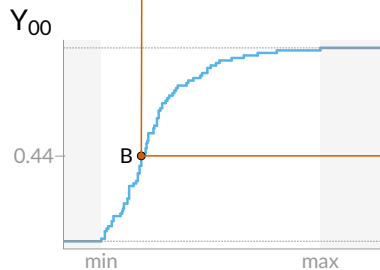
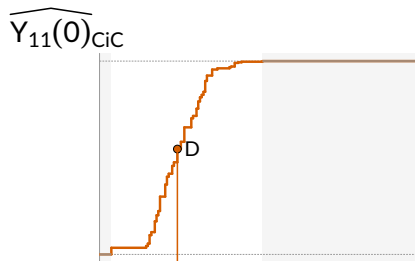
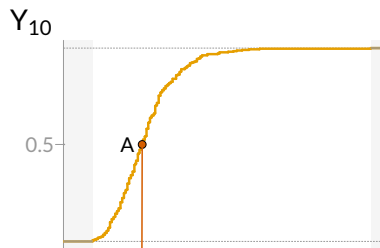
Y_{00}



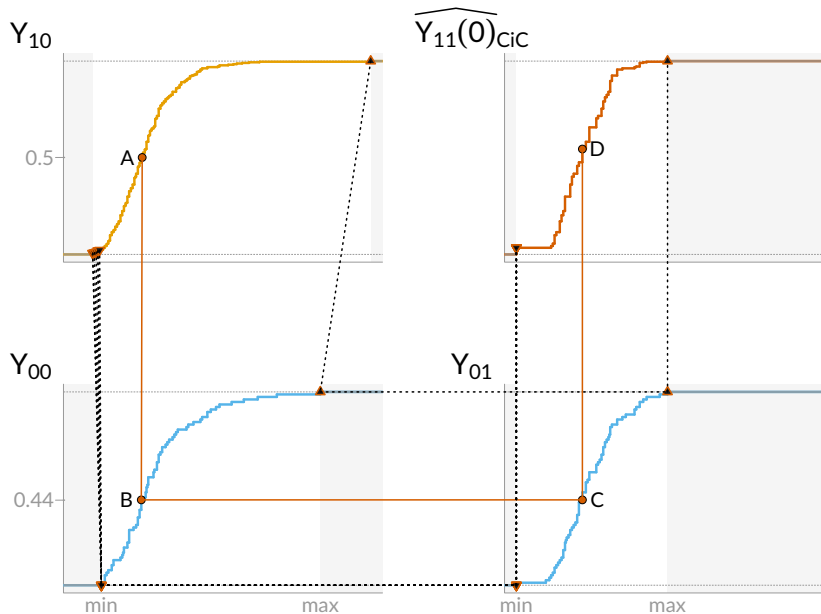
Y_{01}



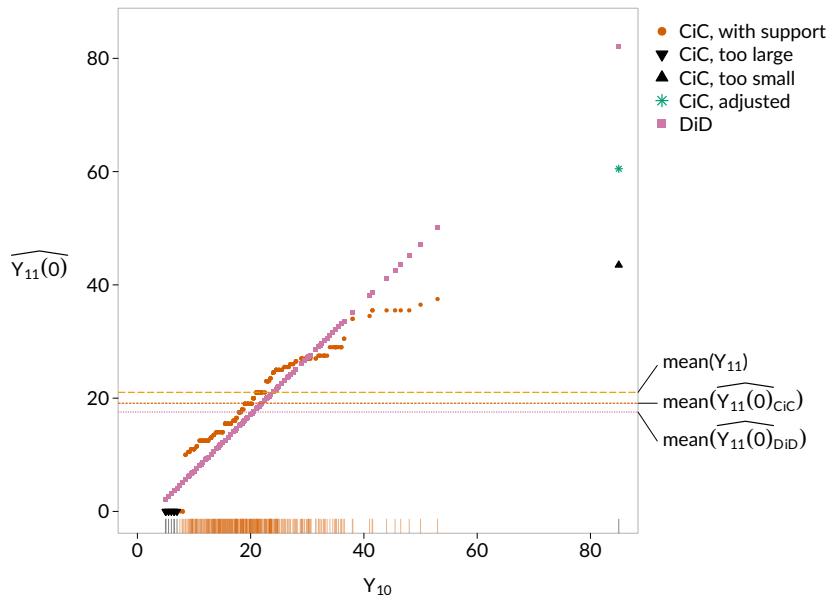
Loop Plot: Changes-in-Changes



Diagonal Lines Show Direction of Support Violation



Imputed Potential Outcomes Plot



Plot demonstrates CiC and DiD methods of extrapolation for observations without support.

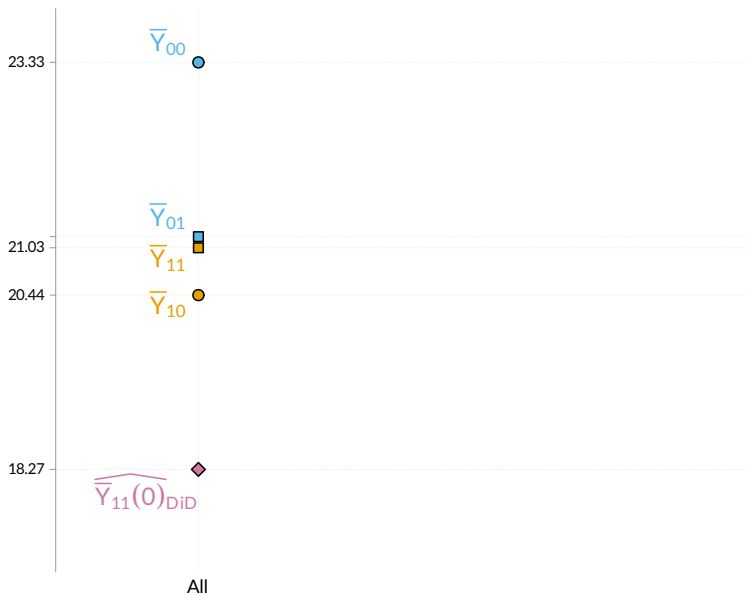
Bounding

- Lack of support implies that extrapolation is necessary for point identification.
- For a lower bound on ATT, we only need to provide an upper bound for the triangle on the right. We use the maximum post-treatment Y value.
- Estimates:

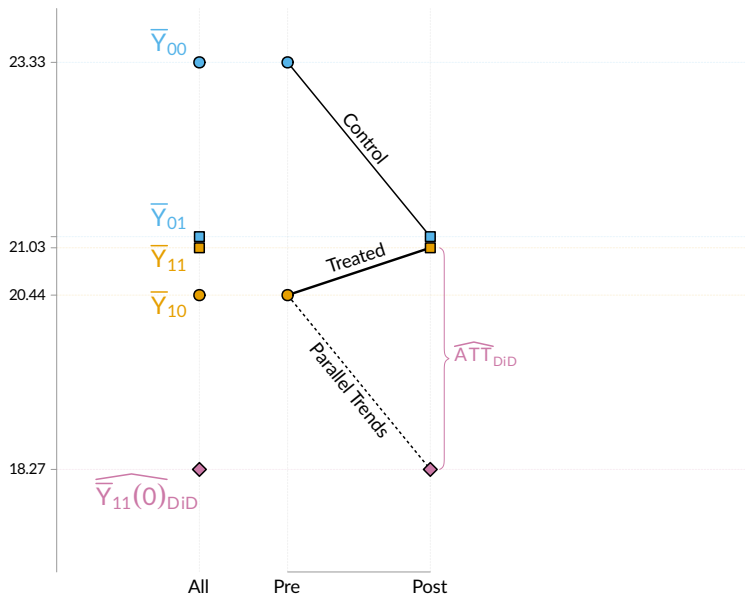
\widehat{ATT}_{DiD}	2.75
\widehat{ATT}_{CiC}	1.92
\widehat{ATT}_{CiC} (lower bound using max $Y_{.1}$)	1.87
\widehat{ATT}_{CiC-r}	1.86

- The balance of the evidence is that the DiD estimate is probably too high.

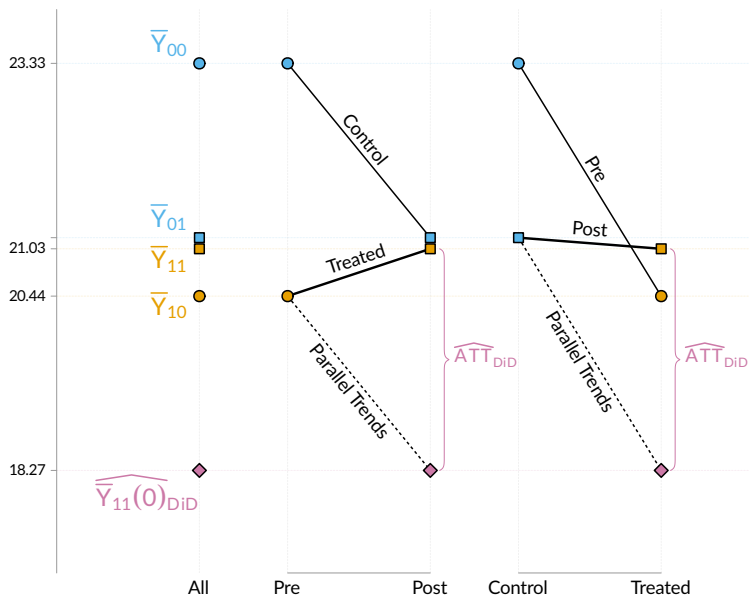
Time vs Group Invariance



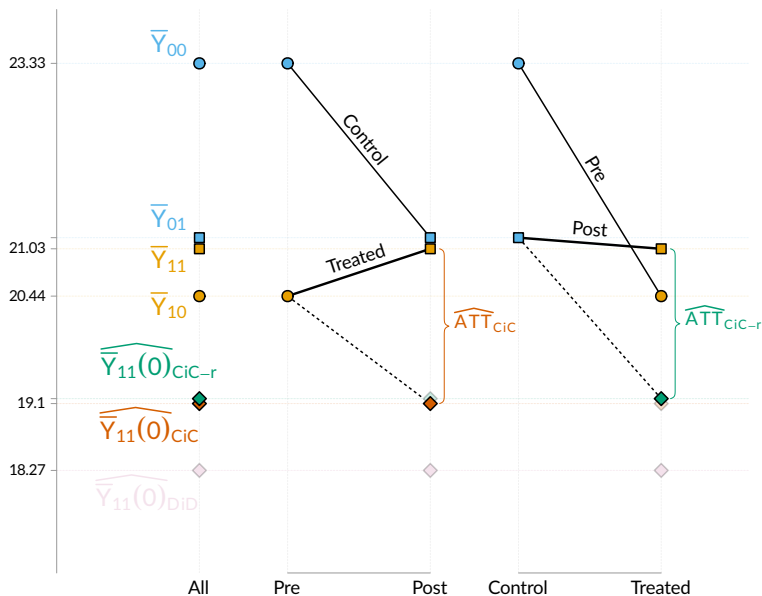
Assume Time Invariant Confounding



Assumption of Group Invariant Confounding Provides Same DiD Estimate



Time and Group Invariance Provide Different CiC Type Estimates



Summary

- Even with only two-time periods, distributional information can reveal issues with DiD, including support problems.
- Loops plots and plots of imputed PO distributions to assess support issues.
- The ATT can be bounded by a support-adjusted CiC estimates. In the minimum wage example CiC estimate provides a lower bound.
- Time-invariant and group-invariant confounding are not equivalent on the quantile scale, simultaneous presentation of CiC and reverse CiC, in addition to support-adjusted estimates for both. In the minimum wage example, CiC-r has no support issues.
- CiC and CiC-r estimates (with and without support adjustments) for the minimum wage example indicate that DiD estimate is likely too high.

Thank you!