

The Impact of Natural Disasters on Education: Evidence from Standardized Testing

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Introduction

I exploit quasi-random variation in natural disaster exposure in the United States to answer two questions:

- ▶ **What is the causal effect of natural disasters on academic achievement as measured by standardized test scores?**
- ▶ What is the role of federal disaster assistance? Which counties apply for assistance?

Why is this important?

Negative effects in education affect earnings potential \implies
Inequality in disaster risk exposure could exacerbate economic inequality

Data

- ▶ **Natural disasters:**
 - ▶ Federal Emergency Management Agency (FEMA) declarations
 - ▶ Storms from the National Weather Service (NWS)
 - ▶ Work in progress: Data on extreme heat
- ▶ **Standardized testing outcomes** from the Stanford Education Data Archive ([Reardon et al., 2021](#)):
 - ▶ Cohort standardized average scores by county in Mathematics & Reading Language Arts (RLA)
 - ▶ Grades 3 through 8 for schoolyears 2008/2009 to 2017/2018
- ▶ **Public Assistance applications and payments** from FEMA

Distribution of mean test scores by subgroup

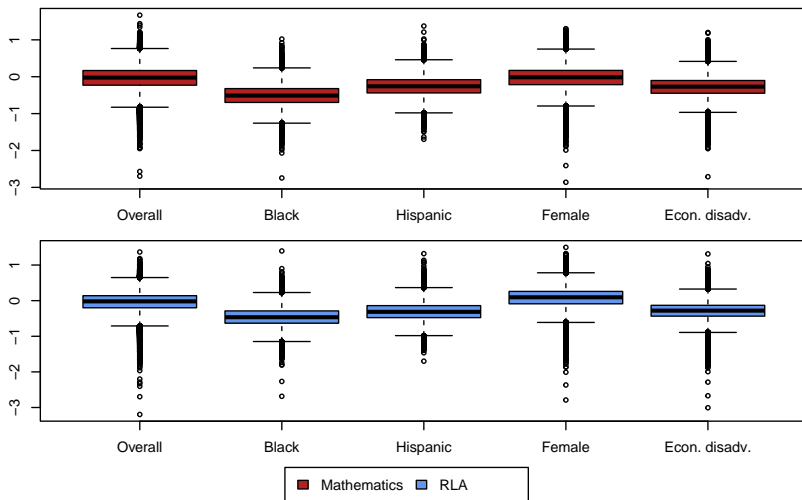


Figure: Boxplots of mean test scores by subgroup

When do counties apply for assistance?

Table: Share of counties that applied for federal assistance following a disaster by disaster type

| | Number of Cases | Applied for Assistance (in %) |
|------------------|-----------------|-------------------------------|
| Dam/Levee Break | 3 | 0.00 |
| Fire | 106 | 10.38 |
| Flood | 85 | 9.41 |
| Hurricane | 1263 | 23.91 |
| Mud/Landslide | 1 | 0.00 |
| Severe Ice Storm | 20 | 0.00 |
| Severe Storm(s) | 154 | 30.52 |
| Tornado | 29 | 79.31 |
| Total | 1661 | 23.54 |

Which counties apply for assistance?

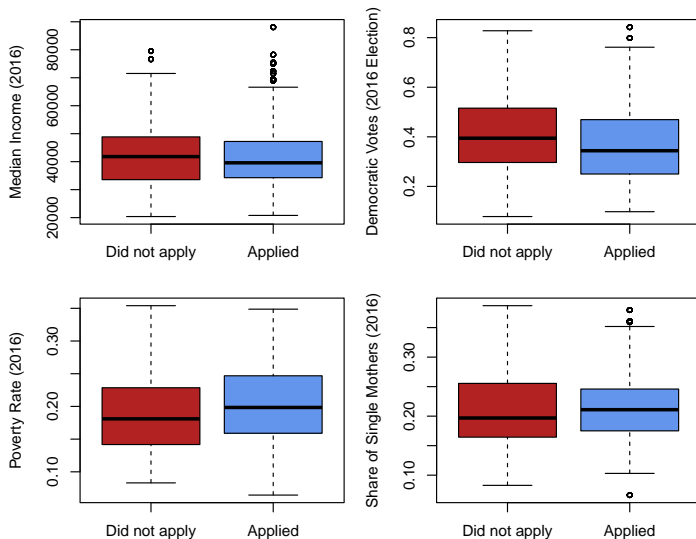


Figure: Boxplots by application status

Empirical Strategy

- ▶ Event-study design:

$$y_{i,t,g} = \beta_{-5}D_{i,t-5} + \sum_{l=-4, l \neq -1}^8 \beta_l D_{i,t-l} + \alpha_i + \lambda_t + \zeta_g + \varepsilon_{i,t,g}$$

- ▶ Treatment begins in the period of first disaster and is absorbing (staggered adoption)
- ▶ But: Always-treated (i.e. disaster in the first year) counties are dropped
- ▶ Never-treated counties act as the baseline

Empirical Strategy: Identification

- ▶ Natural disasters and test scores are plausibly **independent conditional on location** (county fixed-effects)
- ▶ Heterogenous treatment effects \implies simple TWFE is inadequate (de Chaisemartin and D'Haultfœuille, 2020; Sun and Abraham, 2021)
- ▶ Solution: Interaction-Weighted Estimator by Sun and Abraham (2021)
- ▶ Identifying Assumptions: Parallel Trends & No Anticipatory Behavior
- ▶ IW consistently estimates a weighted average of cohort average treatment effects on the treated (CATT)

Main Results: FEMA

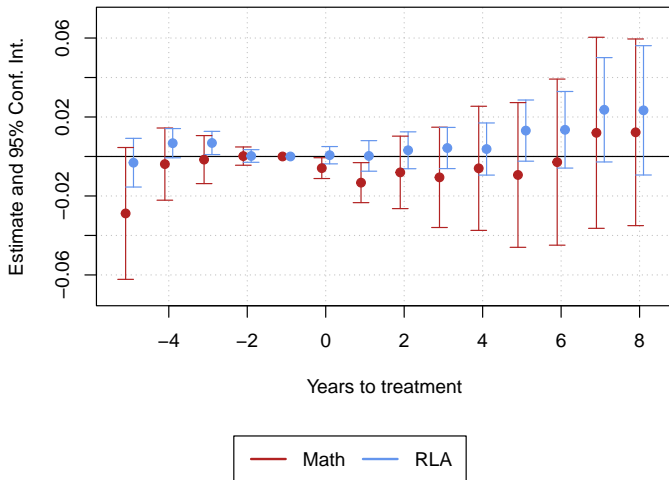


Figure: Dynamic Treatment effects in relative time: FEMA disaster data

Main Results: Subgroups, FEMA

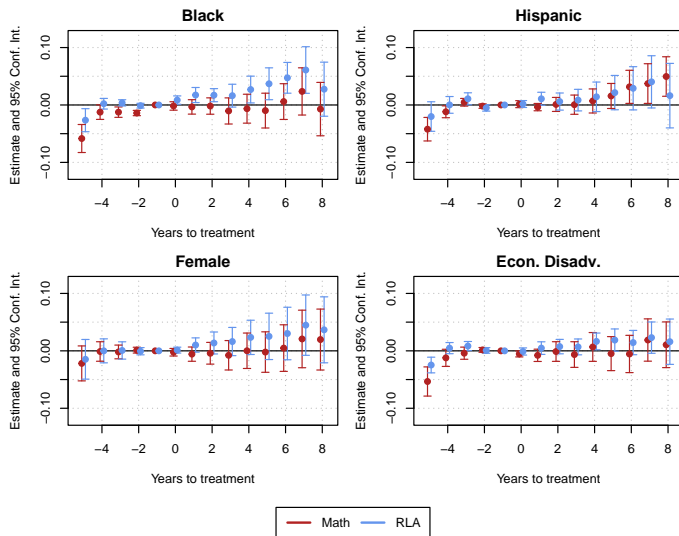


Figure: Dynamic Treatment effects in relative time: FEMA disaster data

Main Results: Storms

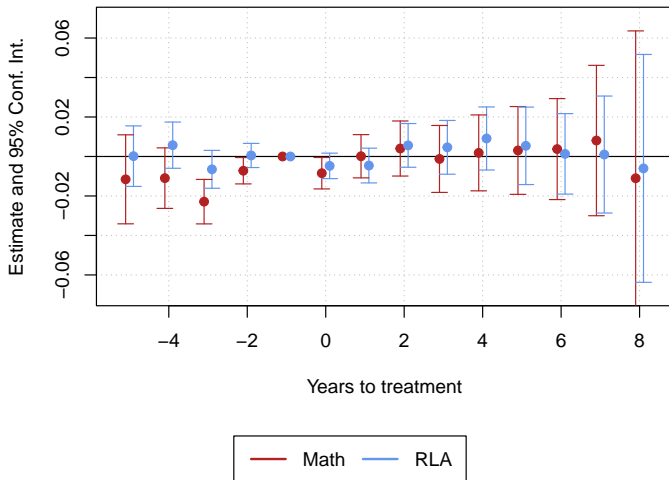


Figure: Dynamic Treatment effects in relative time: NWS storm data

Main Results: Subgroups, Storms

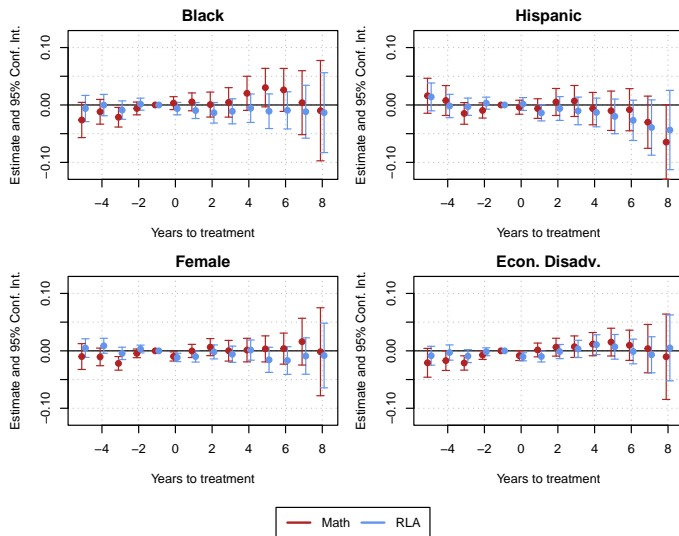


Figure: Dynamic Treatment effects in relative time: NWS storm data

Are these results driven by changes in county composition?

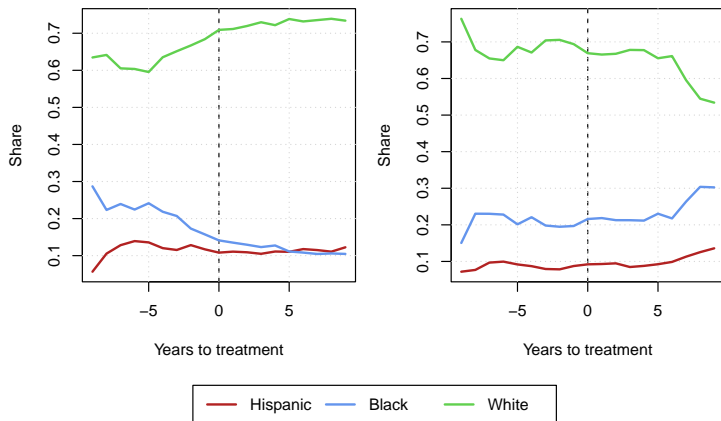


Figure: Aggregated ethnic shares by treatment timing based on FEMA disasters (left) and on NWS storms (right)

References

- de Chaisemartin, C. and D'Haultfœuille, X. (2020). Two-way fixed effects estimators with heterogeneous treatment effects. *American Economic Review*, 110(9):2964–96.
- Reardon, S., Kalogrides, D., Ho, A., Shear, B., Fahle, E., Jang, H., and Chavez, B. (2021). Stanford education data archive (version 4.1).
- Sun, L. and Abraham, S. (2021). Estimating dynamic treatment effects in event studies with heterogeneous treatment effects. *Journal of Econometrics*, 225(2):175–199.