Making the most of the Fragile Families and Child Wellbeing Study

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Annual Meeting of the Population Association of America Slides posted: github.com/ilundberg/slides



Study all causes of some outcome by throwing them all in a model.

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A recommendation

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Aim to answer a specific question that can be stated without a regression model

— A specific question can be descriptive

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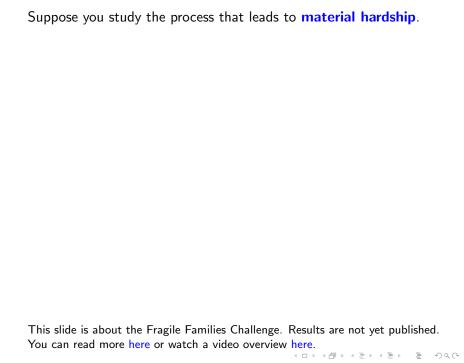
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*Coefficients made up for example.

Black	.05
Hispanic	.07
Mother < HS	.02
Past material hardship	.56
Income	43
Multigenerational household	43
Married parents	43
Mother depressed	.12
Father depressed	.14
Mother unemployed	.67
Father unemployed	.74

Table 1. Predictors of material hardship.

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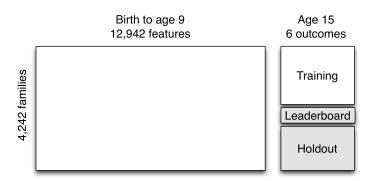
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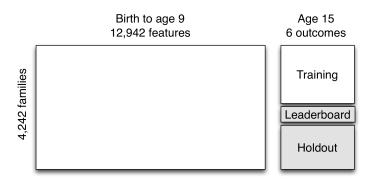
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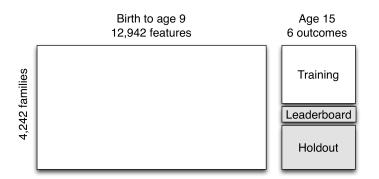
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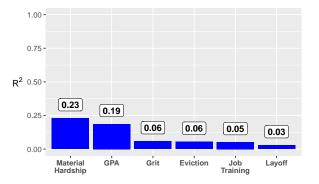
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If you aim to

- predict an outcome well (our goal) or
- understand **all causes** of an outcome (harder)

you are likely to be disappointed with these data.

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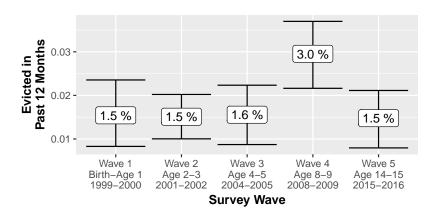
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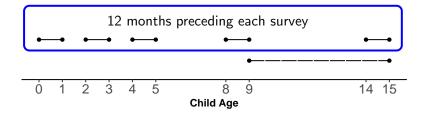
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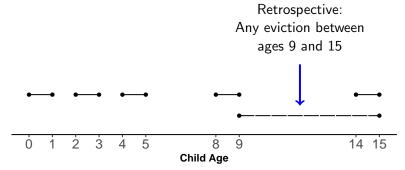
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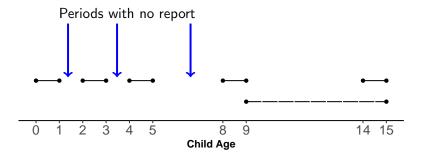
In the past 12 months, were you evicted from your home or apartment for not paying the rent or mortgage?

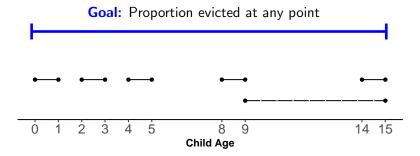


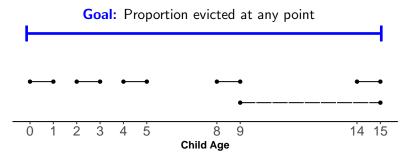




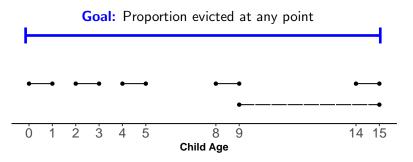








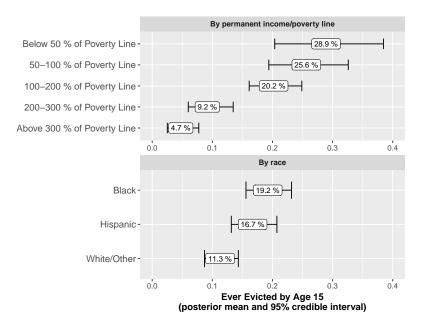
Lower bound: 7.9 % reported an eviction

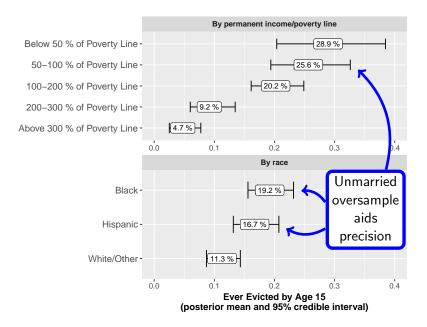


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Model-based: 14.8 % evicted at some point

→ Imputes over periods with no report





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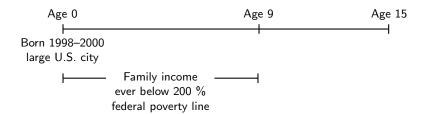


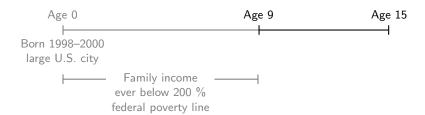
Does **public housing** protect low-income families from **eviction**?

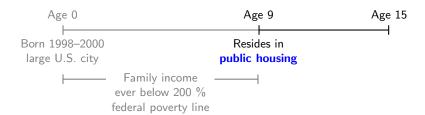
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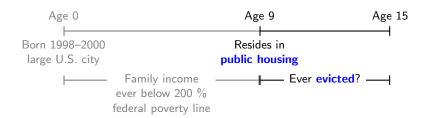


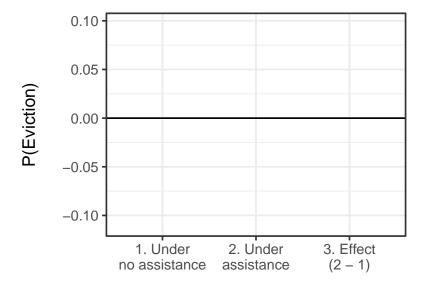


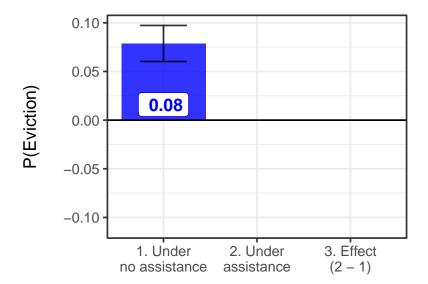


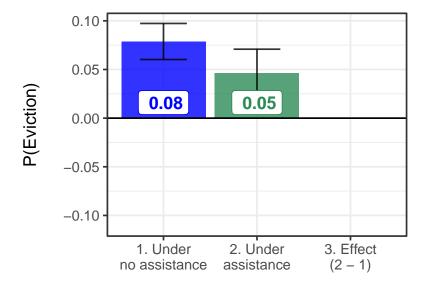


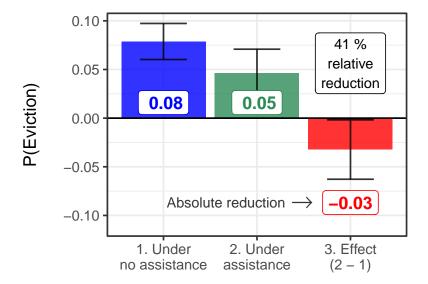












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- Not Nationally representative
- Not Residents of cities
- Yes Probability sample of births in U.S. cities with populations over 200,000 in 1998–2000

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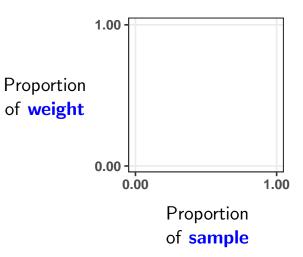
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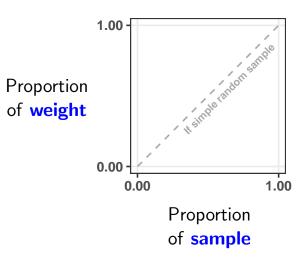
Tendency toward disadvantage is a **consequence** of this.

Weights are highly unequal.

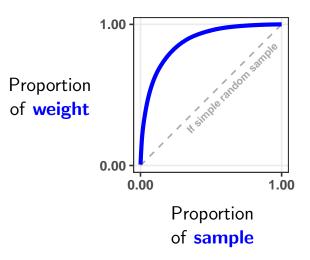
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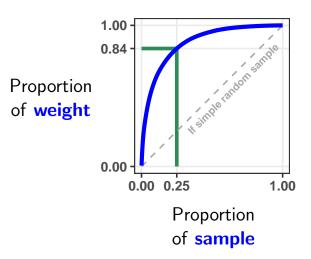
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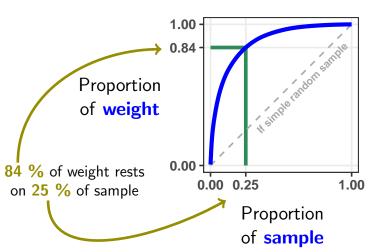
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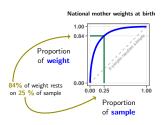


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Weights are not bad.

They are just high-variance.

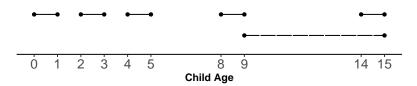
This is a consequence of the oversample that enables precise estimates for non-marital births.

Advice: Think carefully about whether to weight.

Weighted and unweighted samples are very different.

Comparison to other datasets

Gaps between surveys: Not annual



Target population is specific. (but so is NLSY-CYA and PSID-CDS)

Many likert-type scales influenced by psychology

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Targ^{et} population: births in cities

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For the right question, these are **advantages**

Slides: github.com/ilundberg/slides

References

Lundberg, Ian, and Louis Donnelly. 2019. "A research note on the prevalence of housing eviction among children born in U.S. cities." *Demography* 56(1):391-404. [accepted manuscript][replication package].

Lundberg, Ian, Sarah L. Gold, Louis Donnelly, Jeanne Brooks-Gunn, and Sara S. McLanahan. "Government assistance protects low-income families from eviction and rent non-payment." To be presented at the 2019 Annual Meeting of the Population Association of America.

Salganik, Matthew J., Ian Lundberg, Alex Kindel, Sara S. McLanahan, and over 100 additional authors. "Measuring the predictability of life outcomes with a scientific mass collaboration." Manuscript in progress.

Read more here or watch a video overview here.

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