

Retiring Adult: New Datasets for Fair Machine Learning

John Miller

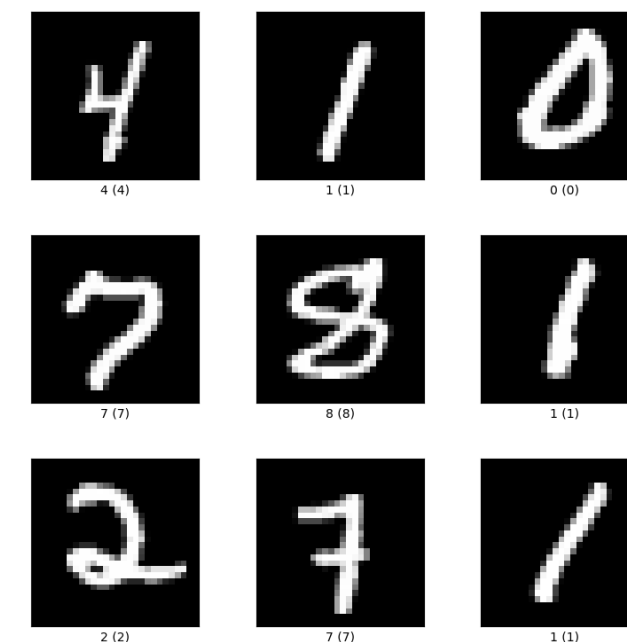
Joint work with Frances Ding, Moritz Hardt, Ludwig Schmidt



<https://folktables.org/>



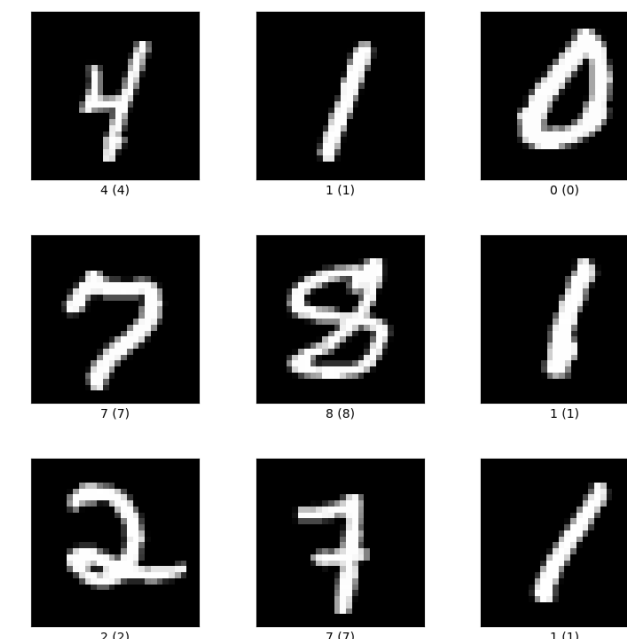
Datasets as benchmarks



kaggle

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- Common training and test sets for model builders



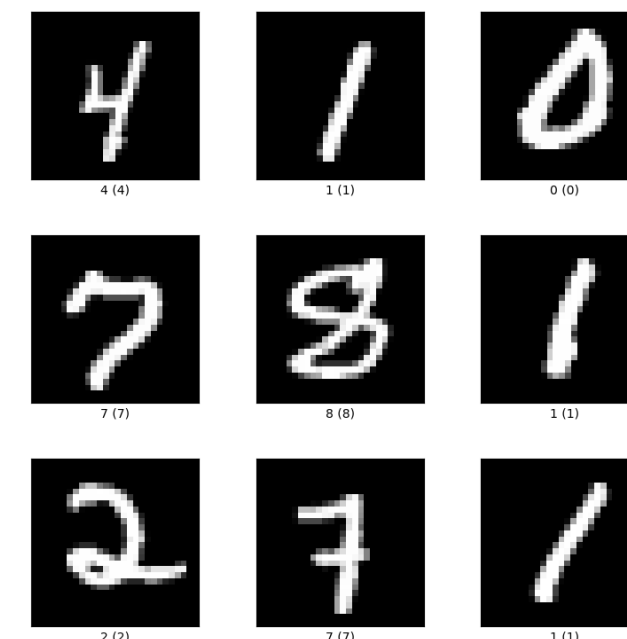
IMAGENET



kaggle

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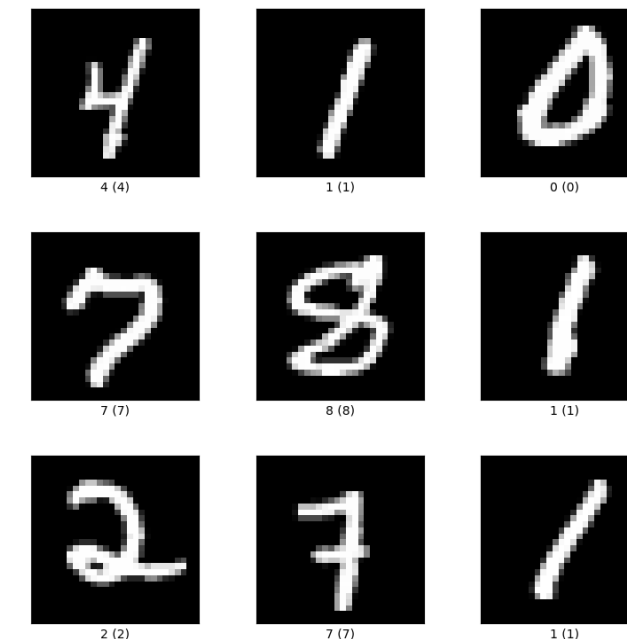
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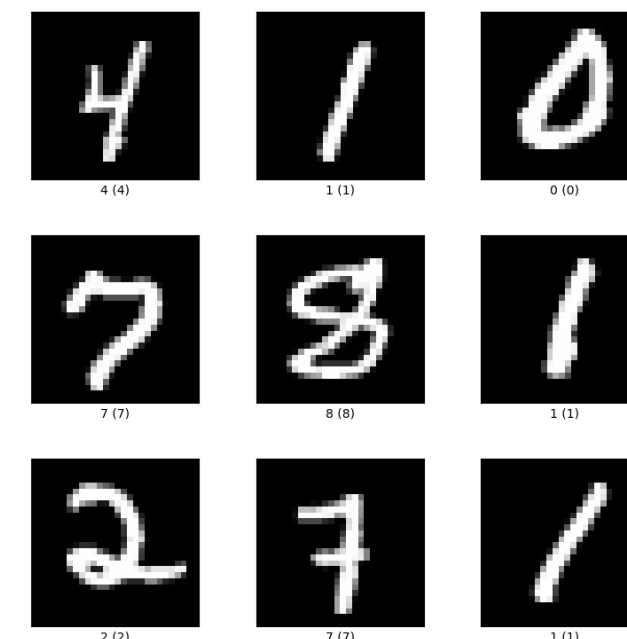
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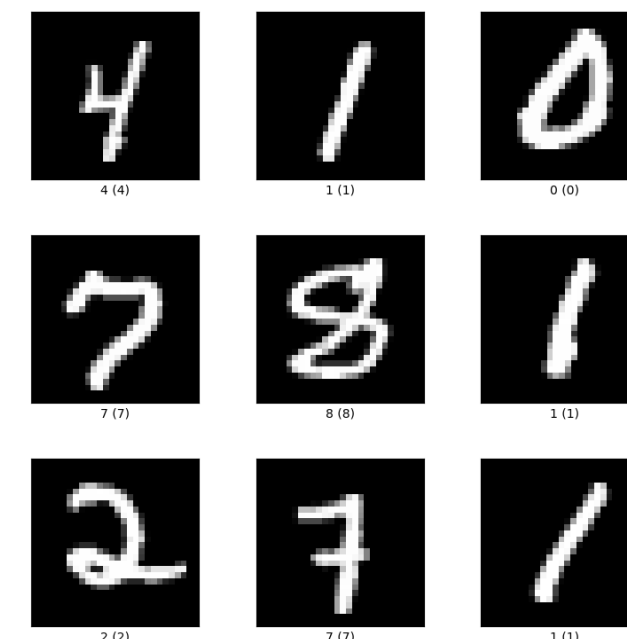
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 - Serve as an interface between academia and industry



Fairness in machine learning

Fairness in machine learning

- Very active research area

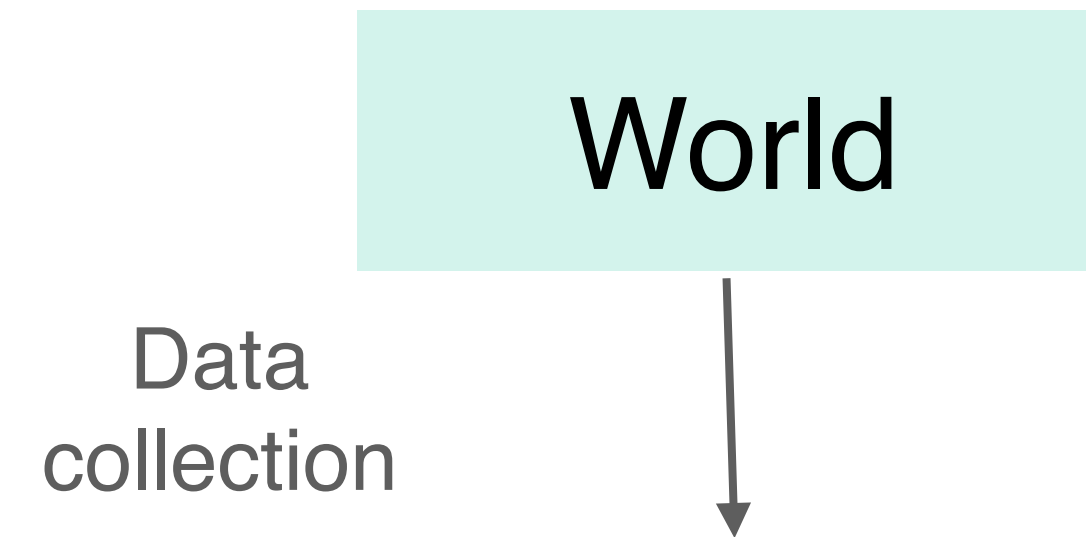
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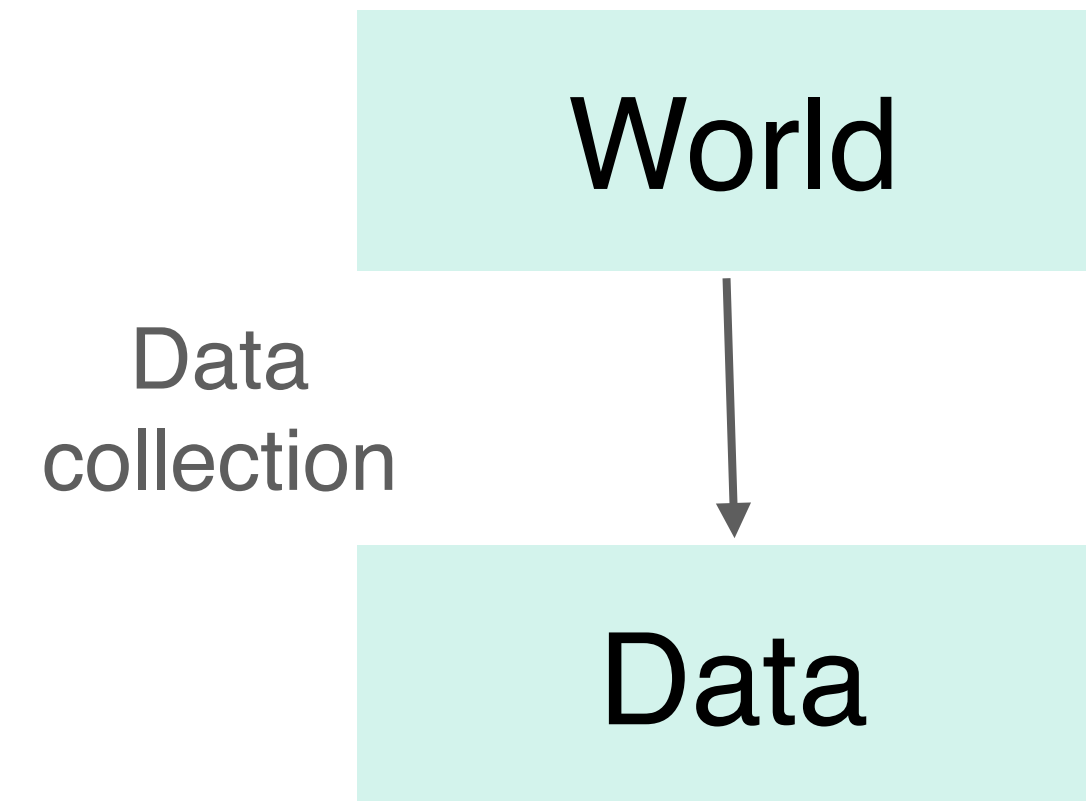
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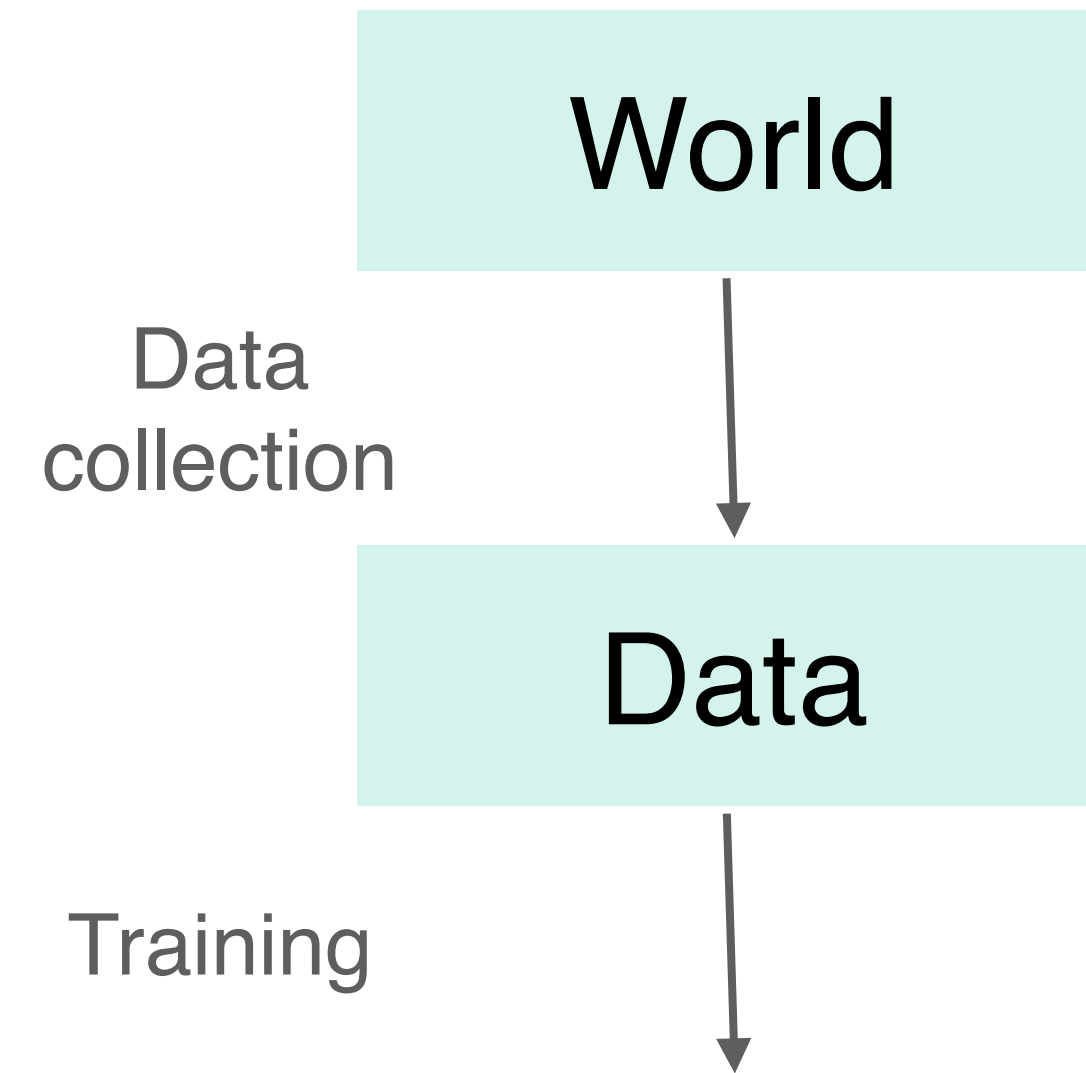
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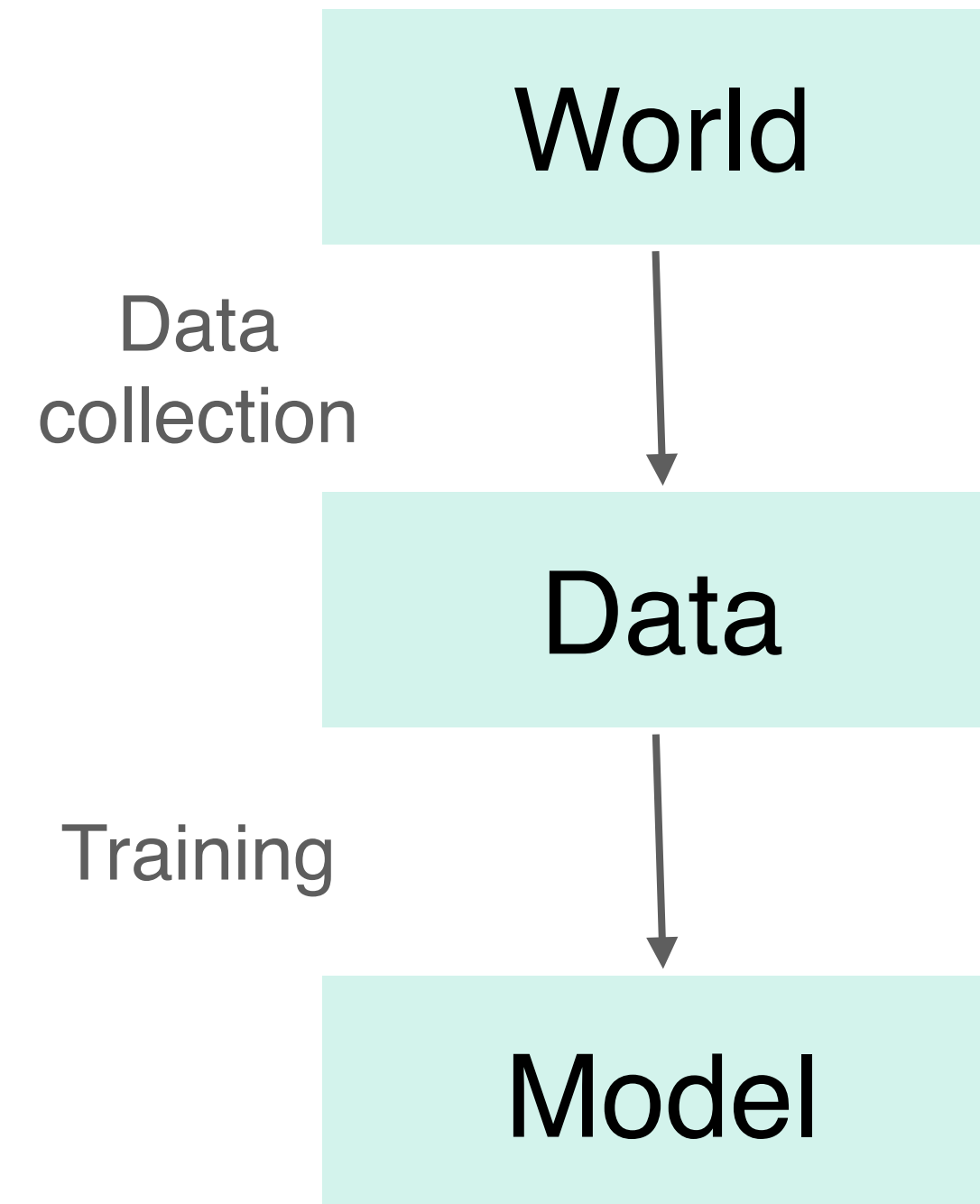
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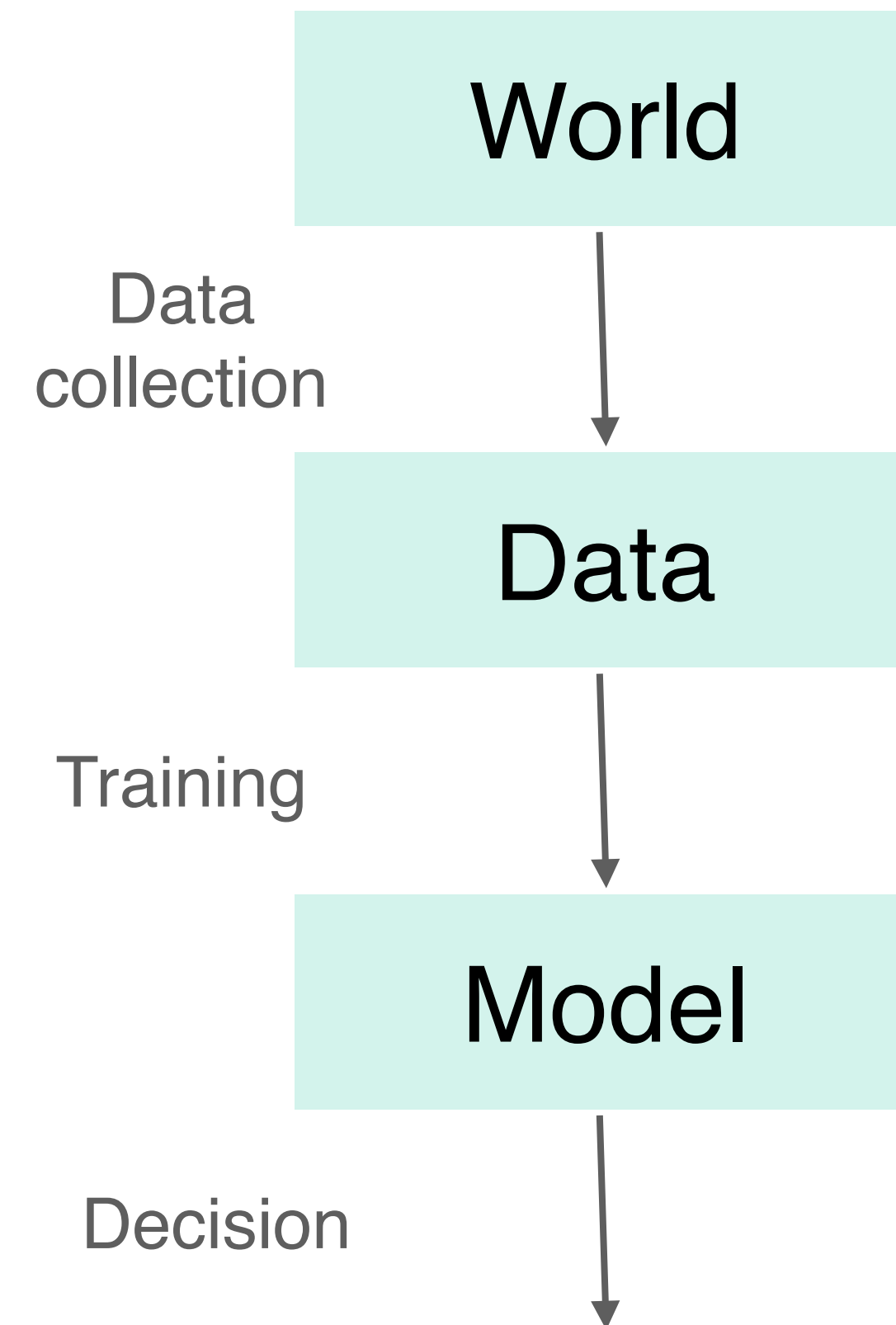
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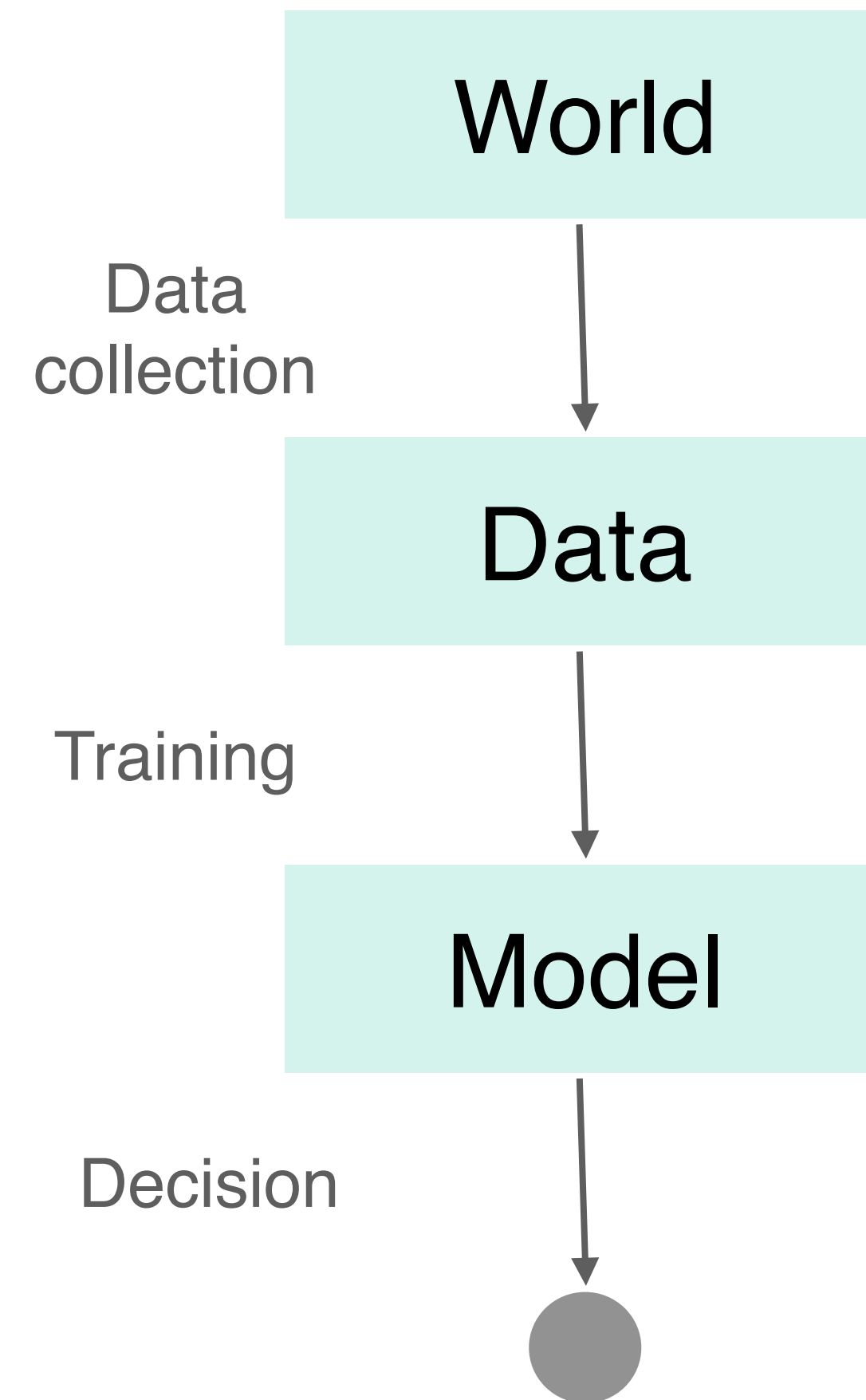
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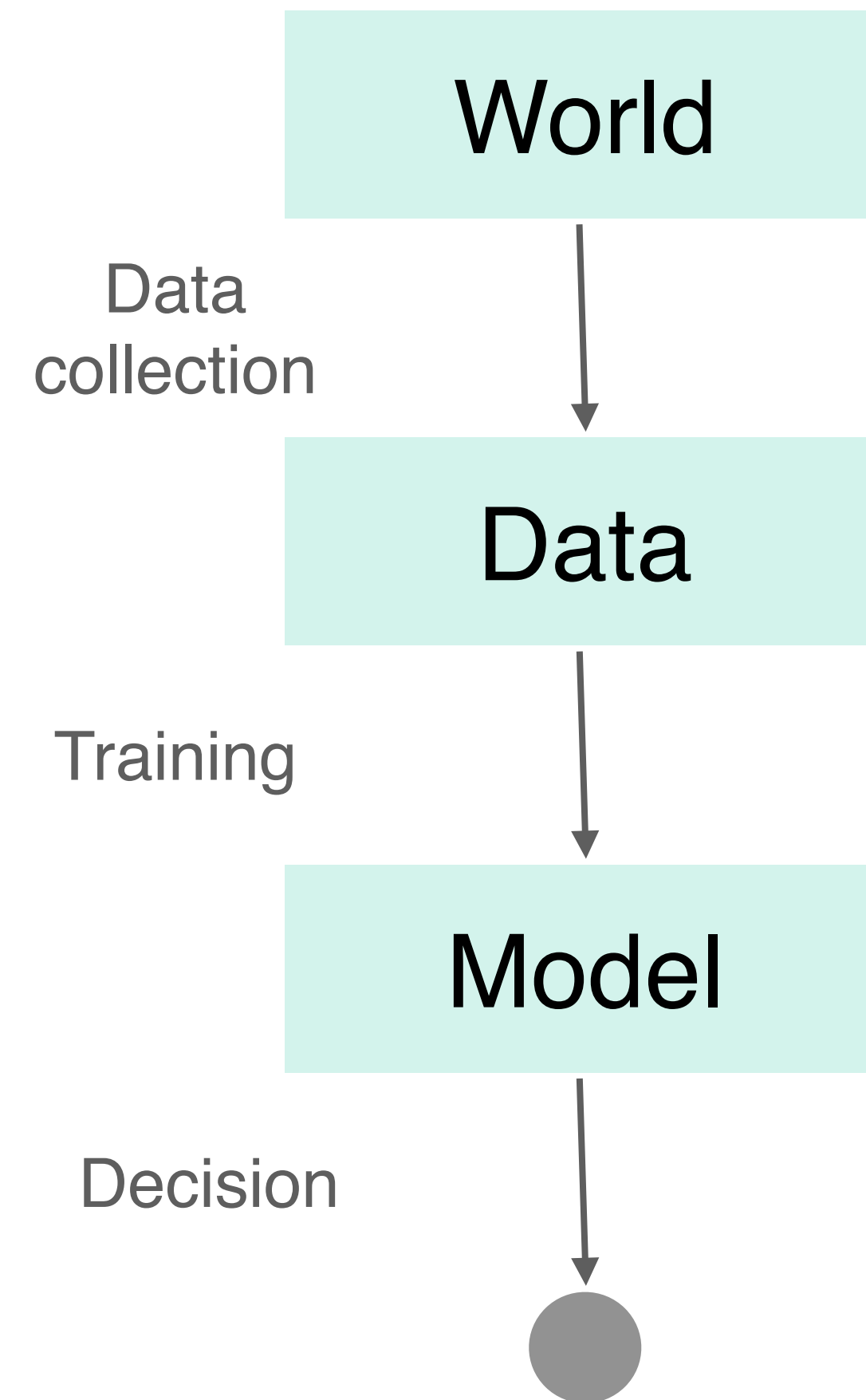
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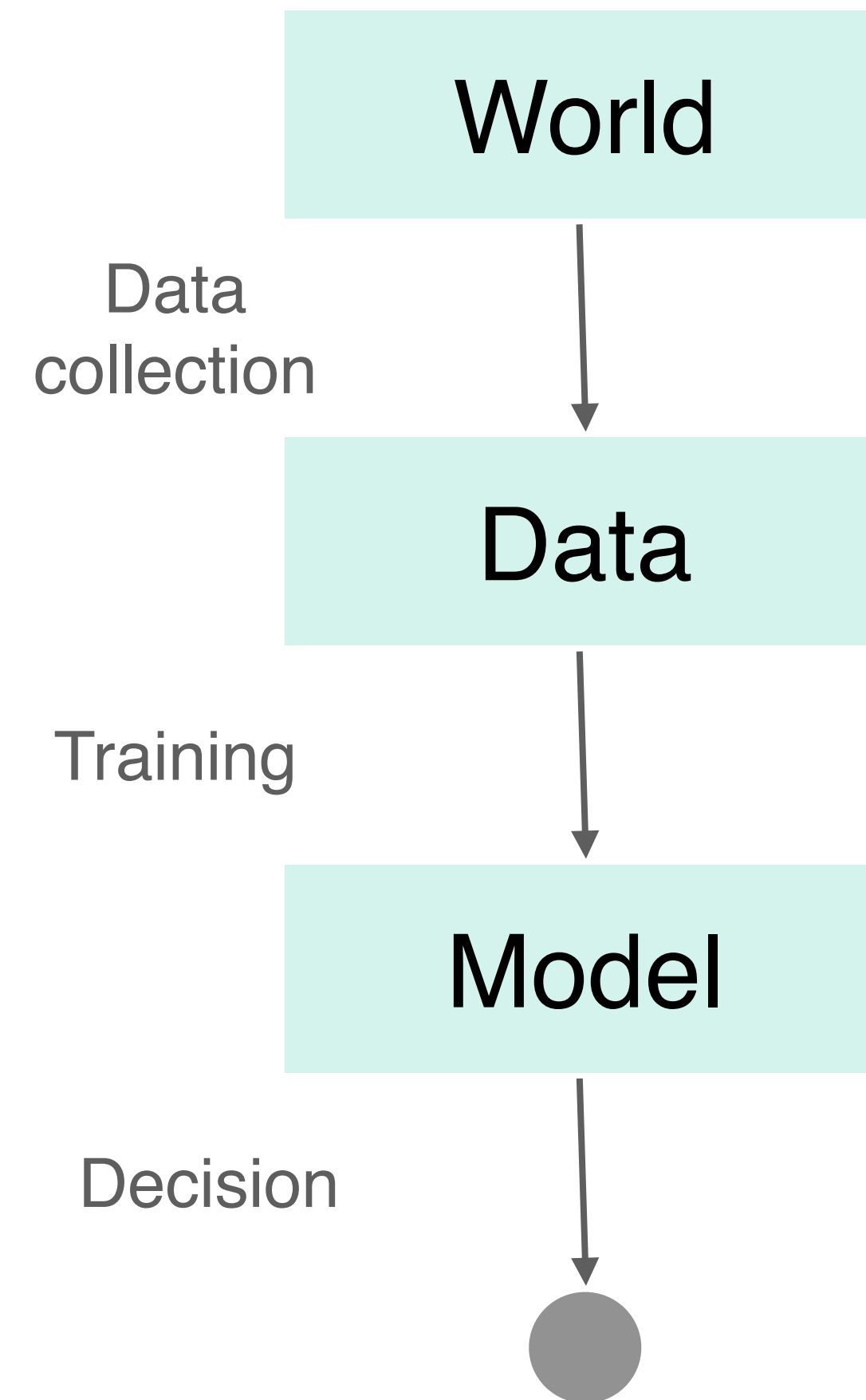
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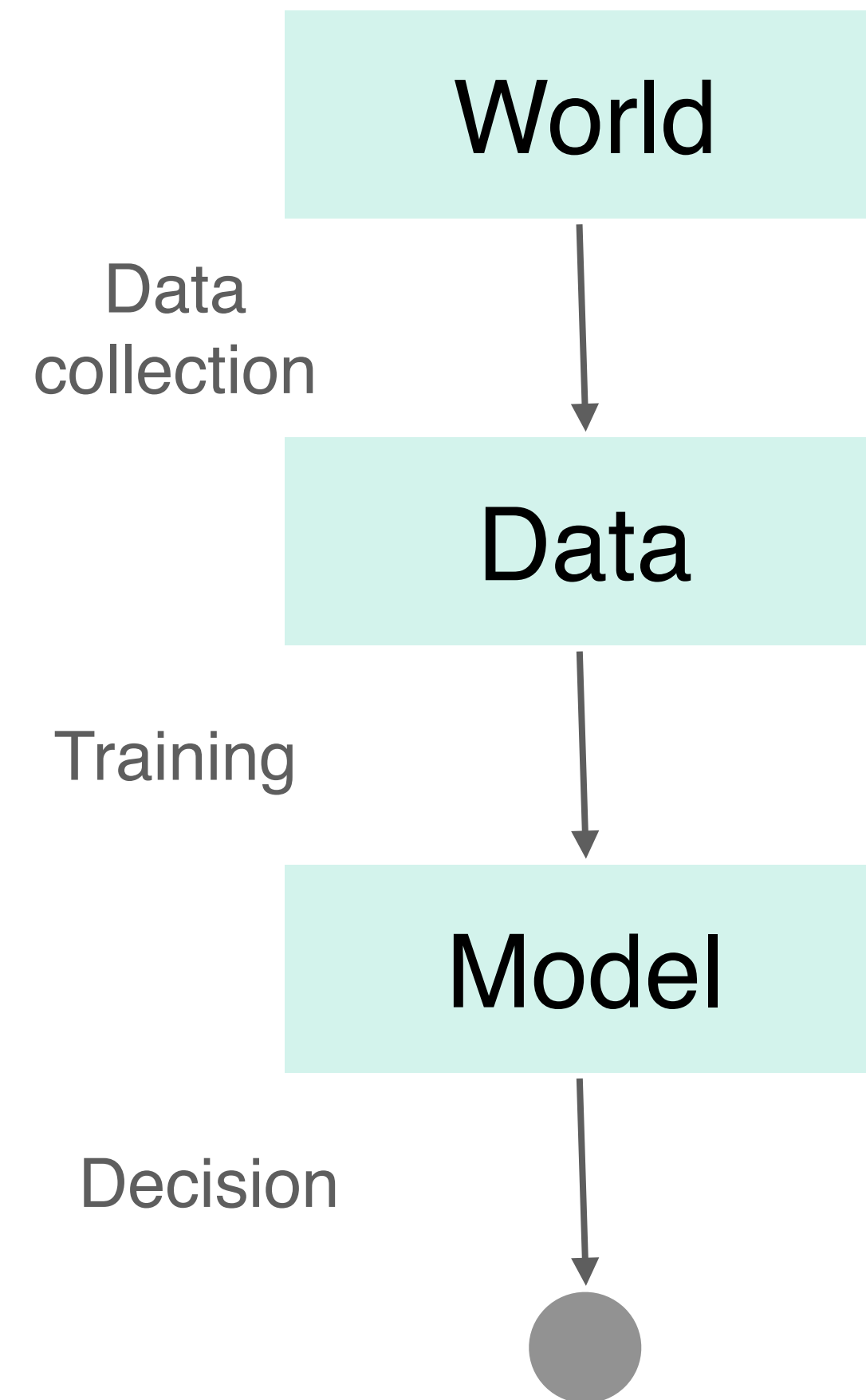
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- When it comes to tabular data, most papers focus on **UCI Adult**




UCI Adult dataset

UCI Adult dataset

← → ↻ 🔒 archive.ics.uci.edu/ml/index.php



Most Popular Data Sets (hits since 2007):

- 4225263:  [Iris](#)
- 2263856:  [Adult](#)
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UCI



Machine Learning Repository

Center for Machine Learning and Intelligent Systems

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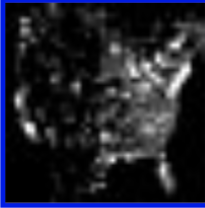



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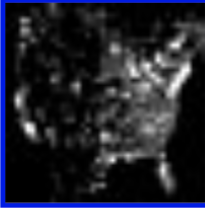


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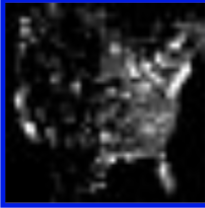

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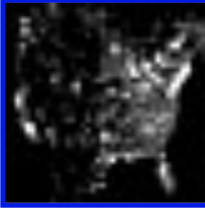
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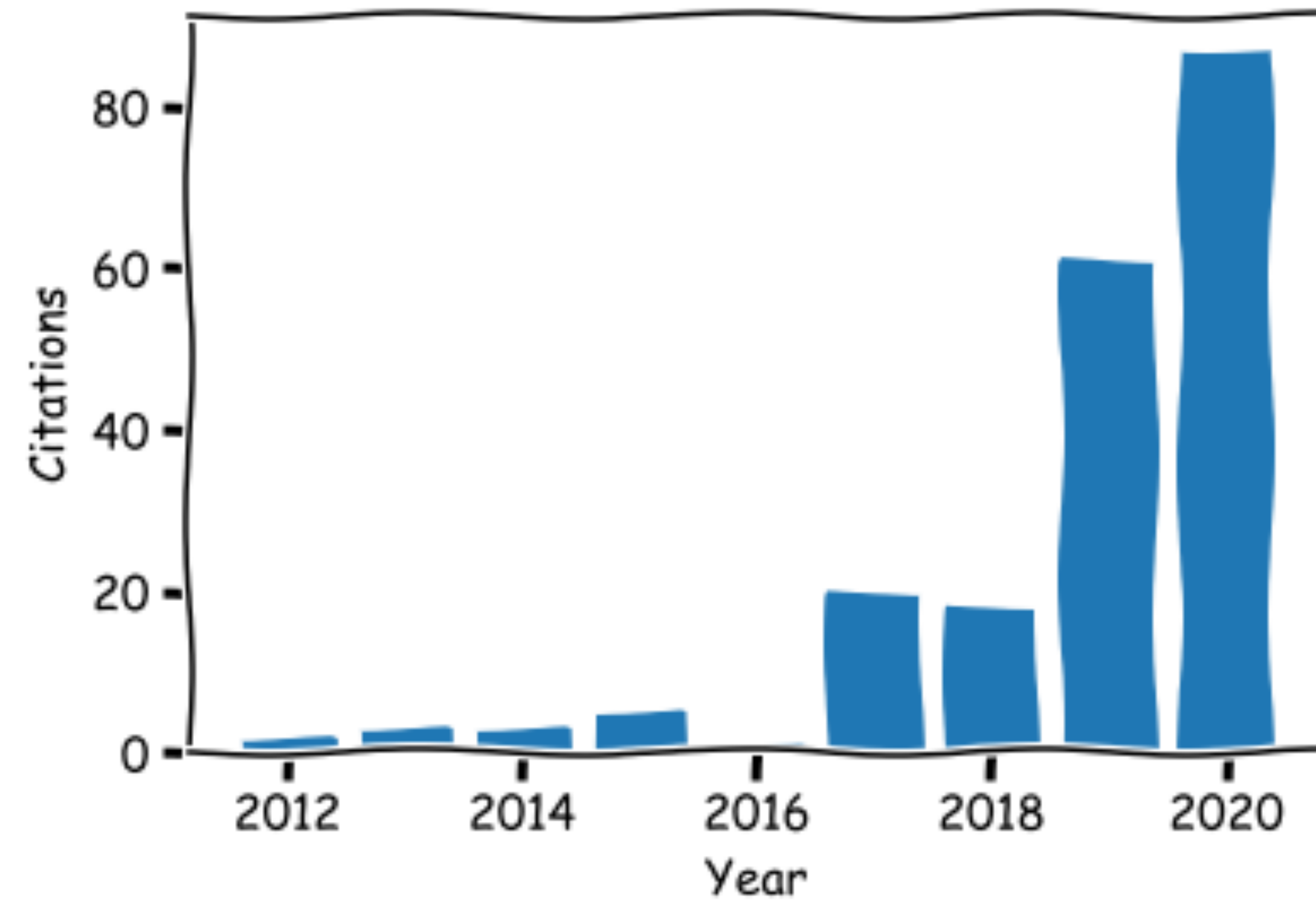
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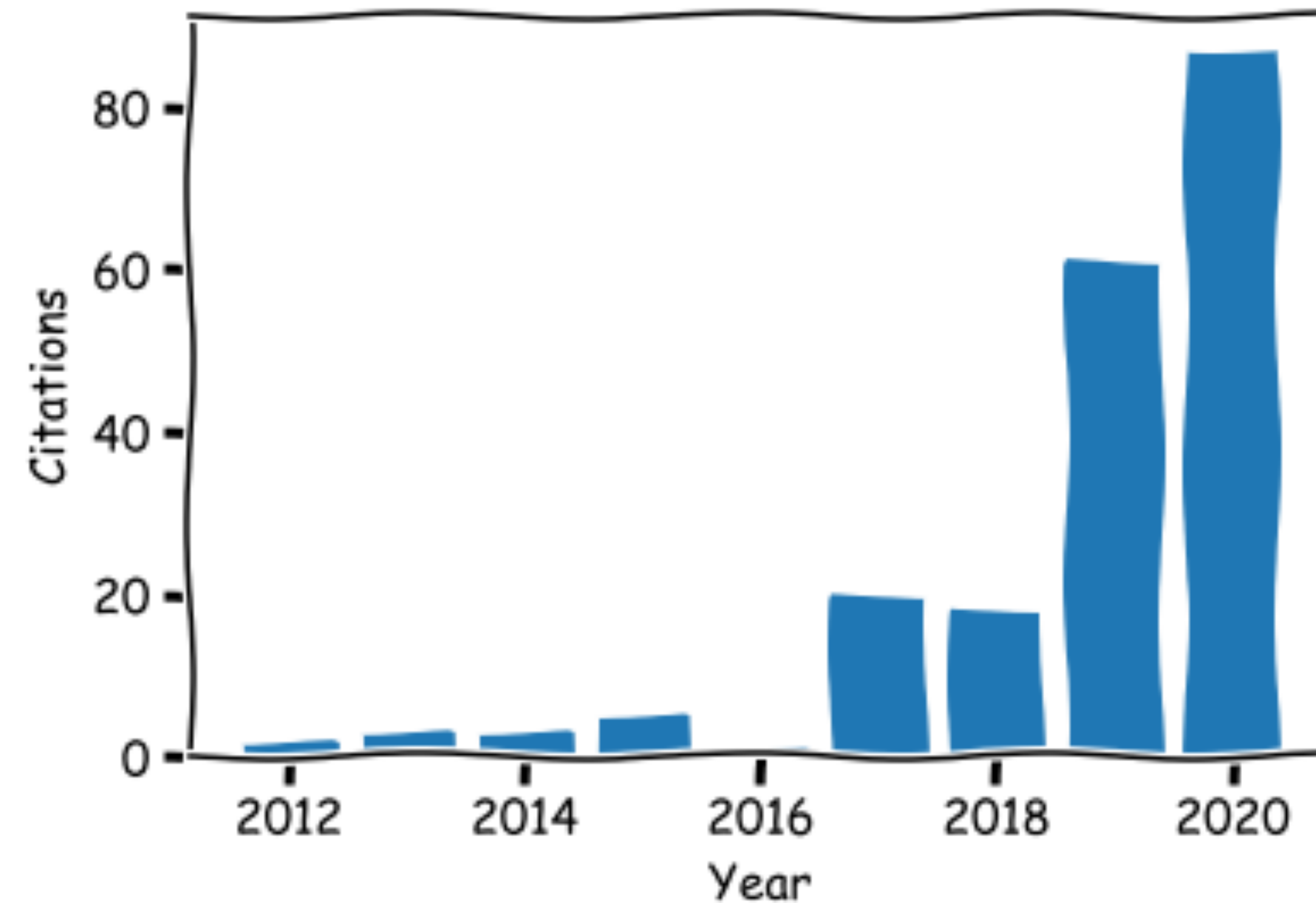
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Fairness papers using UCI Adult

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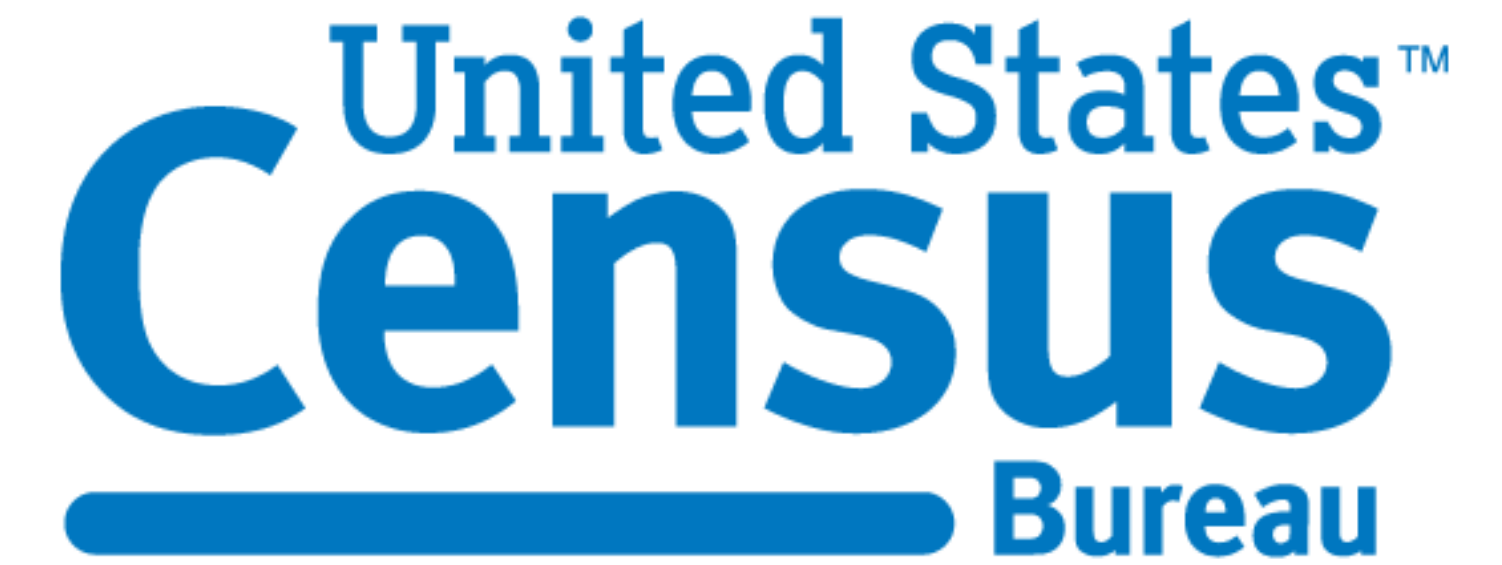
Fairness papers using UCI Adult



Lower bound: 300+ fairness papers cite UCI Adult

***Where does UCI Adult come from?
Is UCI Adult a good empirical foundation for fair ML?***

Reconstruction of UCI Adult



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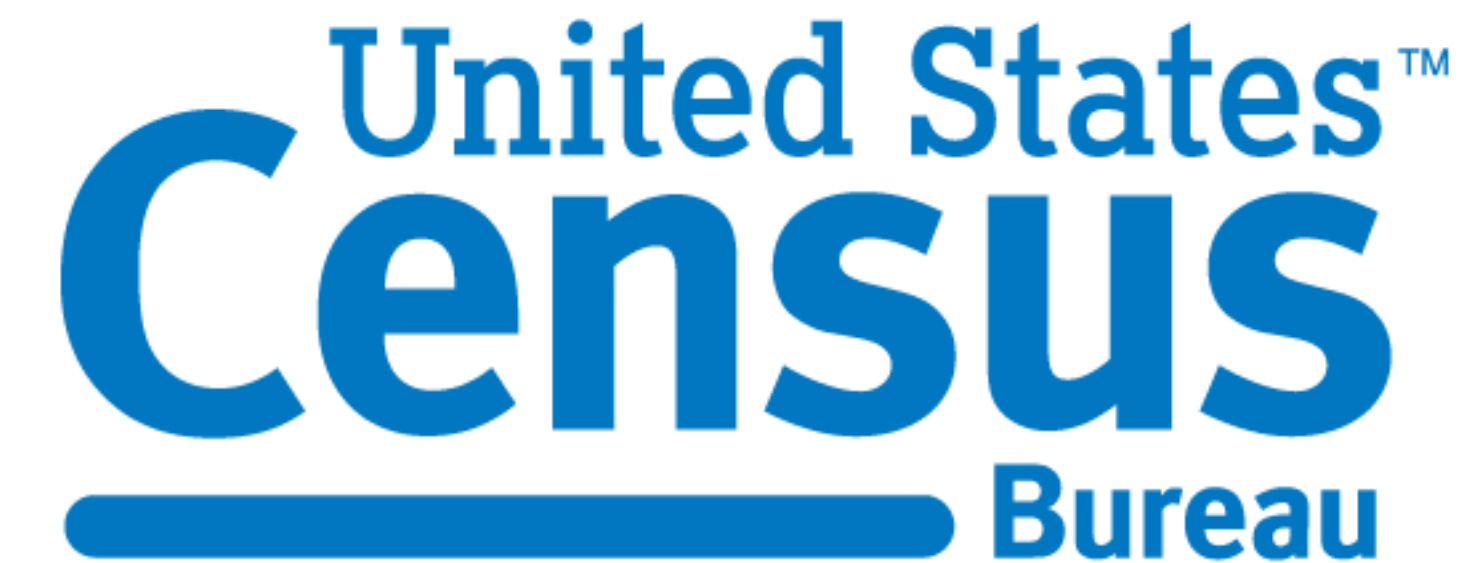
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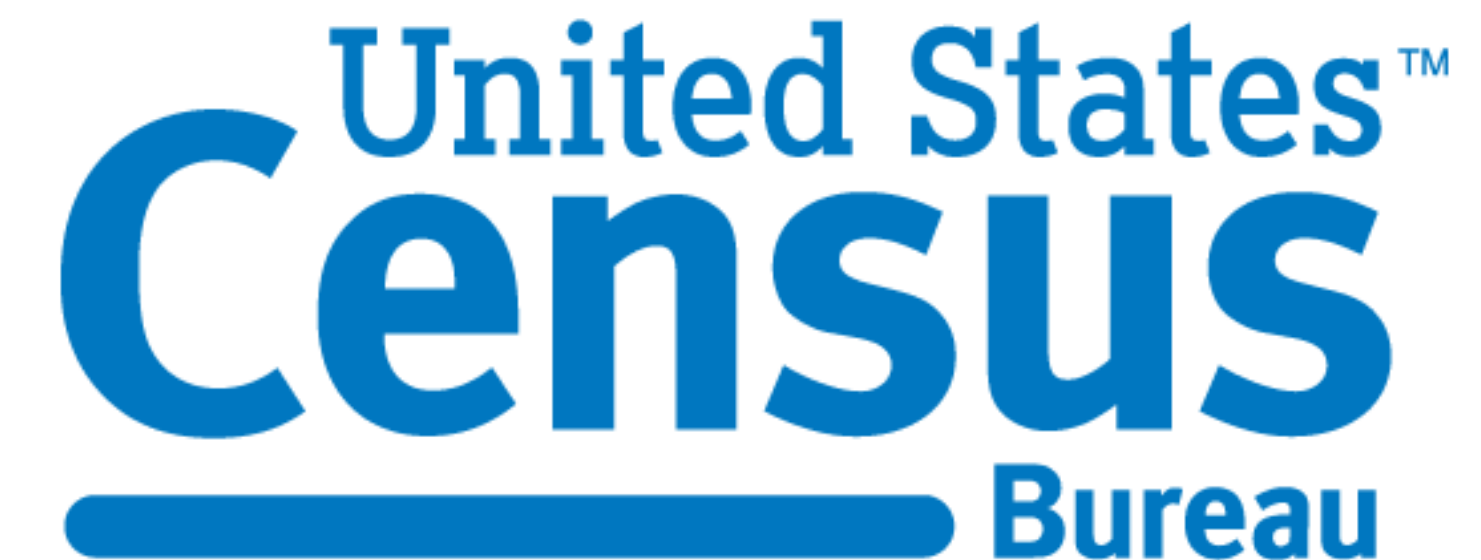
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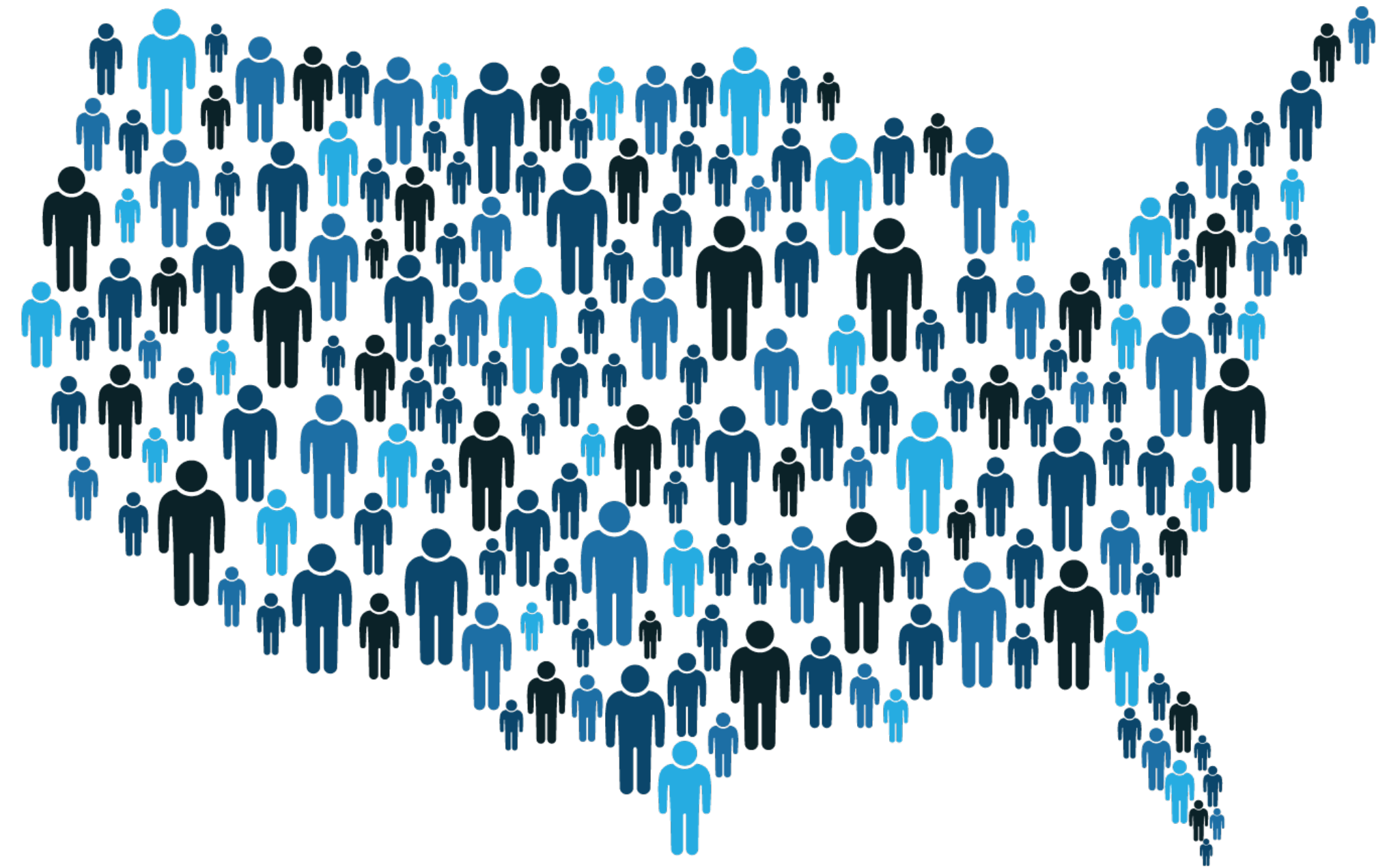


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 - Quality sampling frame capturing the US population
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 - Thorough documentation



Issues with UCI Adult: external validity

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CENSUS '90



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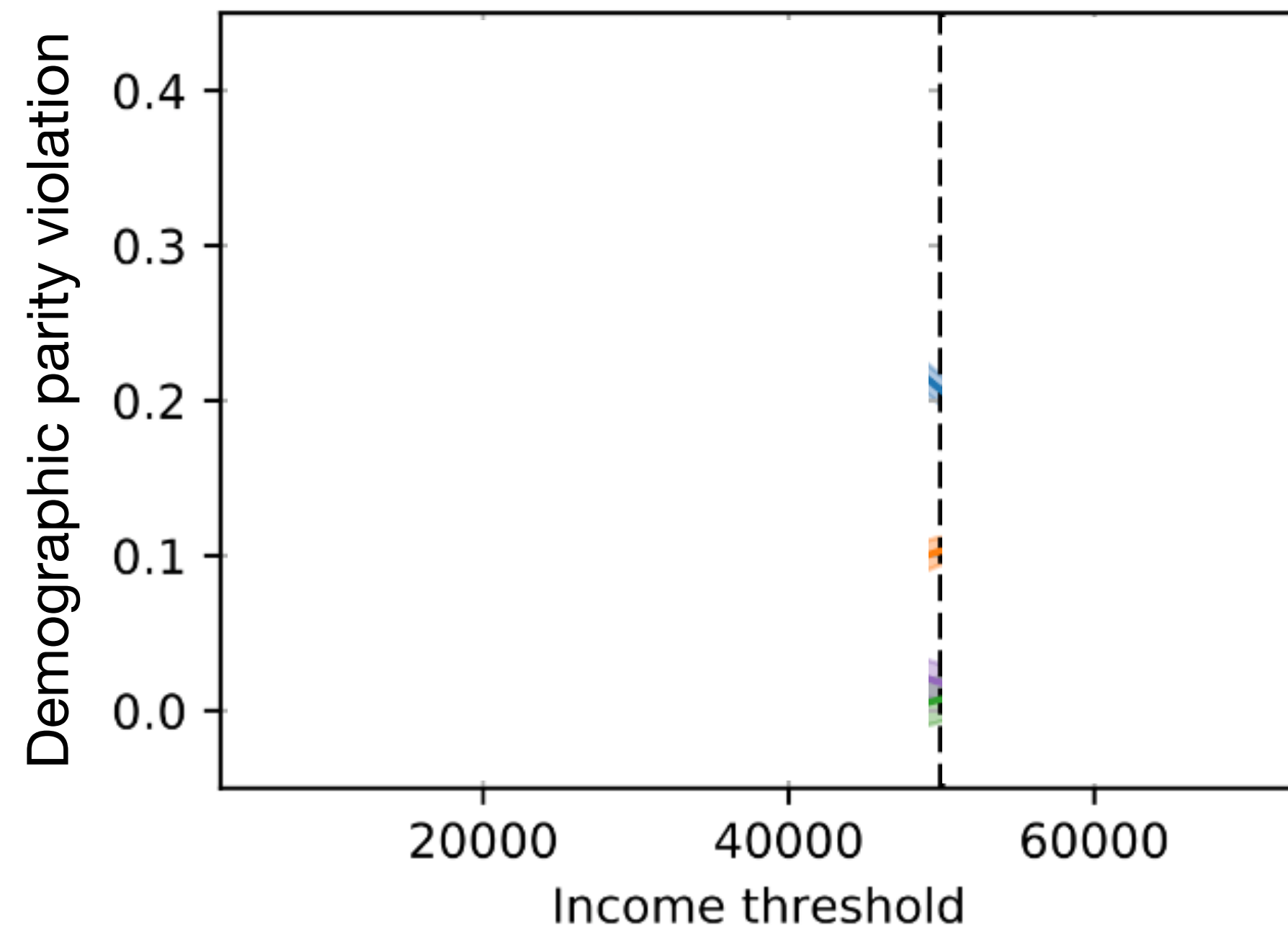
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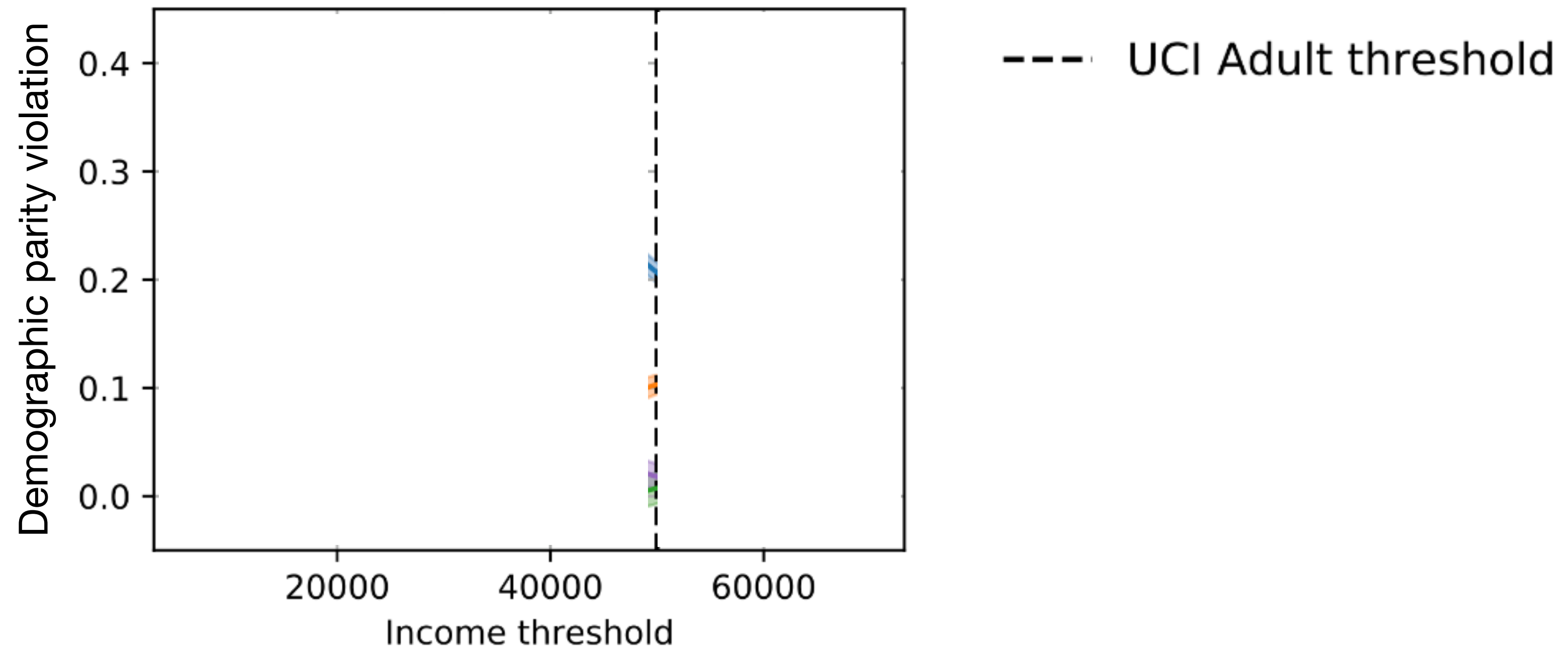
Classifier accuracy is **higher** for individuals in minority groups,
the **opposite** setting of much of fair ML

Varying the income threshold

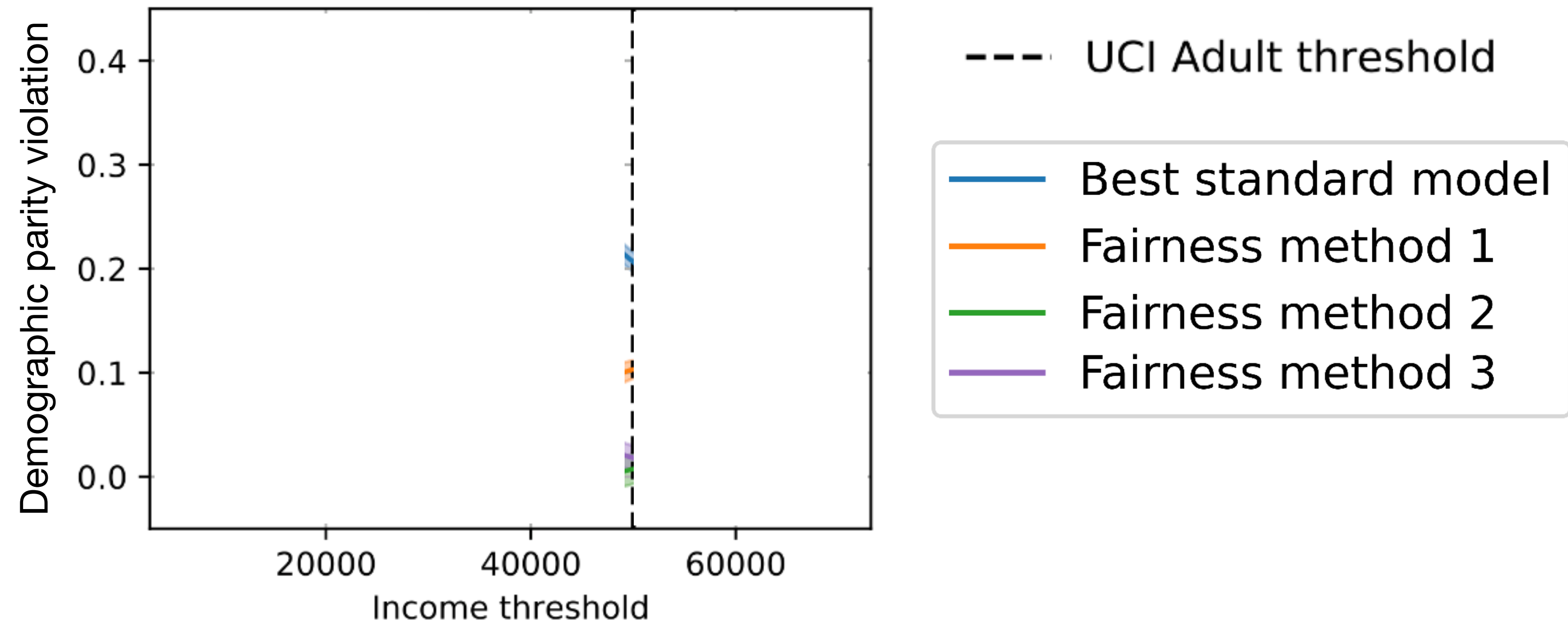
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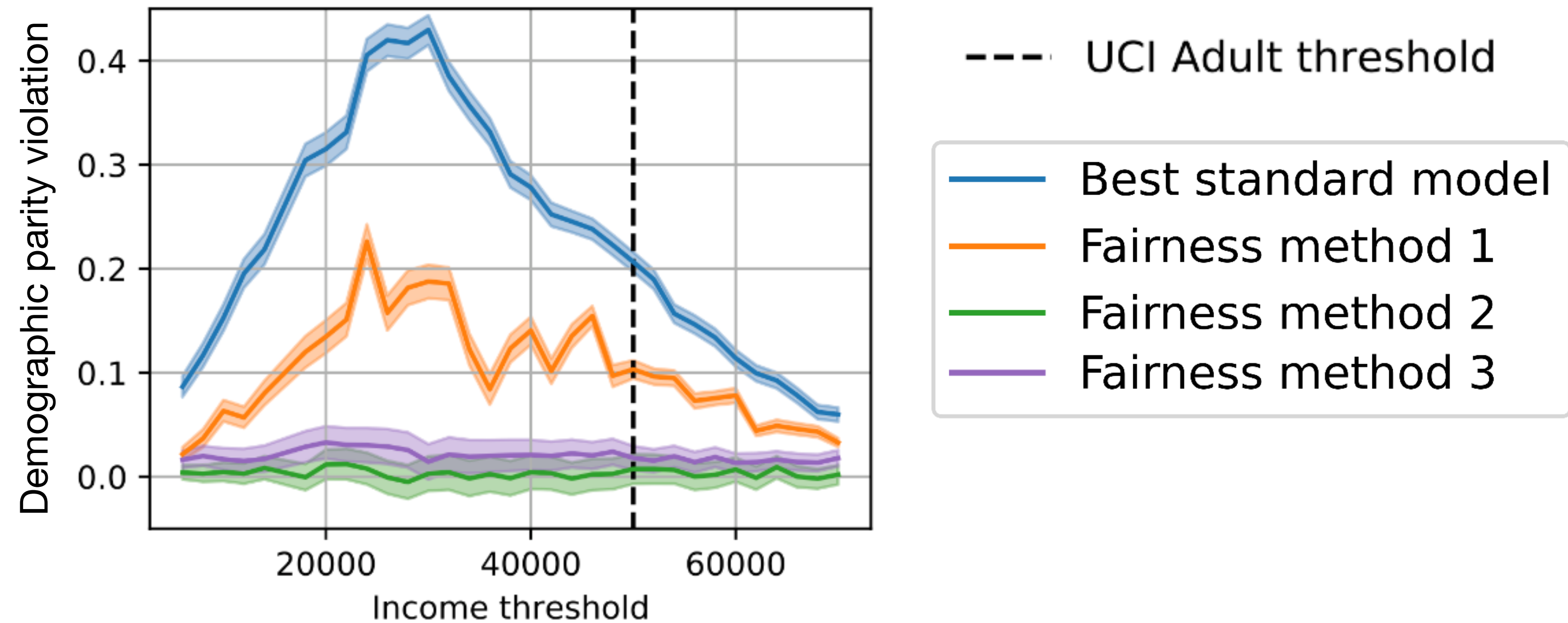
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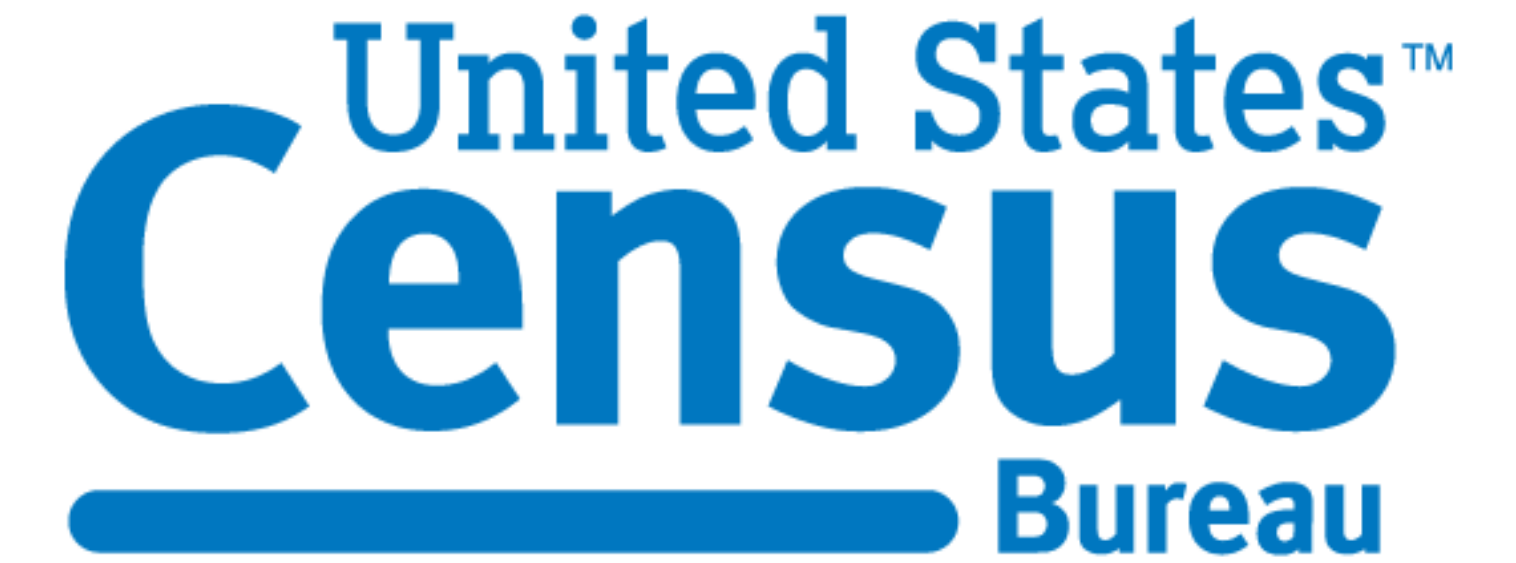


Varying the income threshold



***Can we get the best of Census data while avoiding
issues with UCI Adult?***

Opportunity for new datasets



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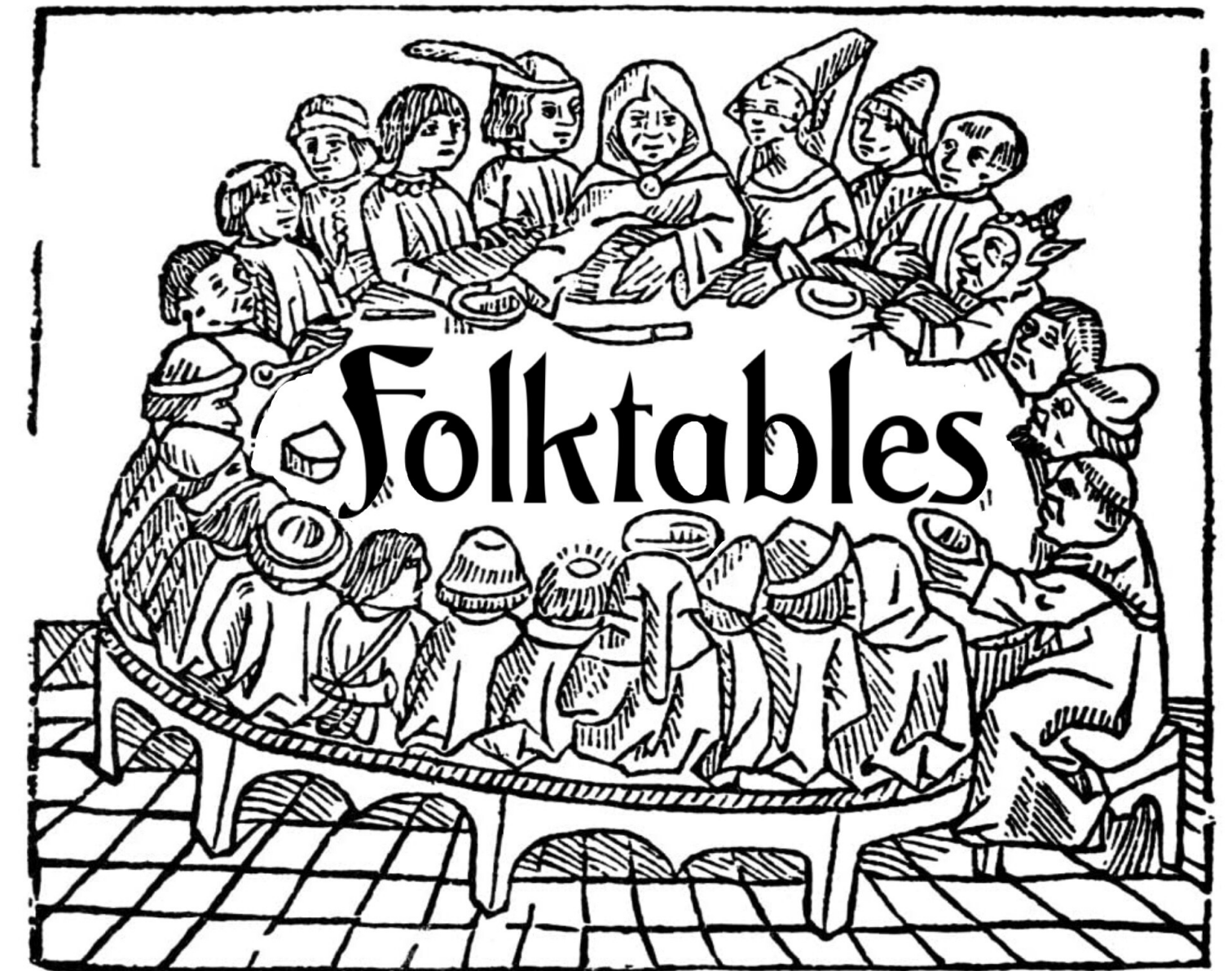


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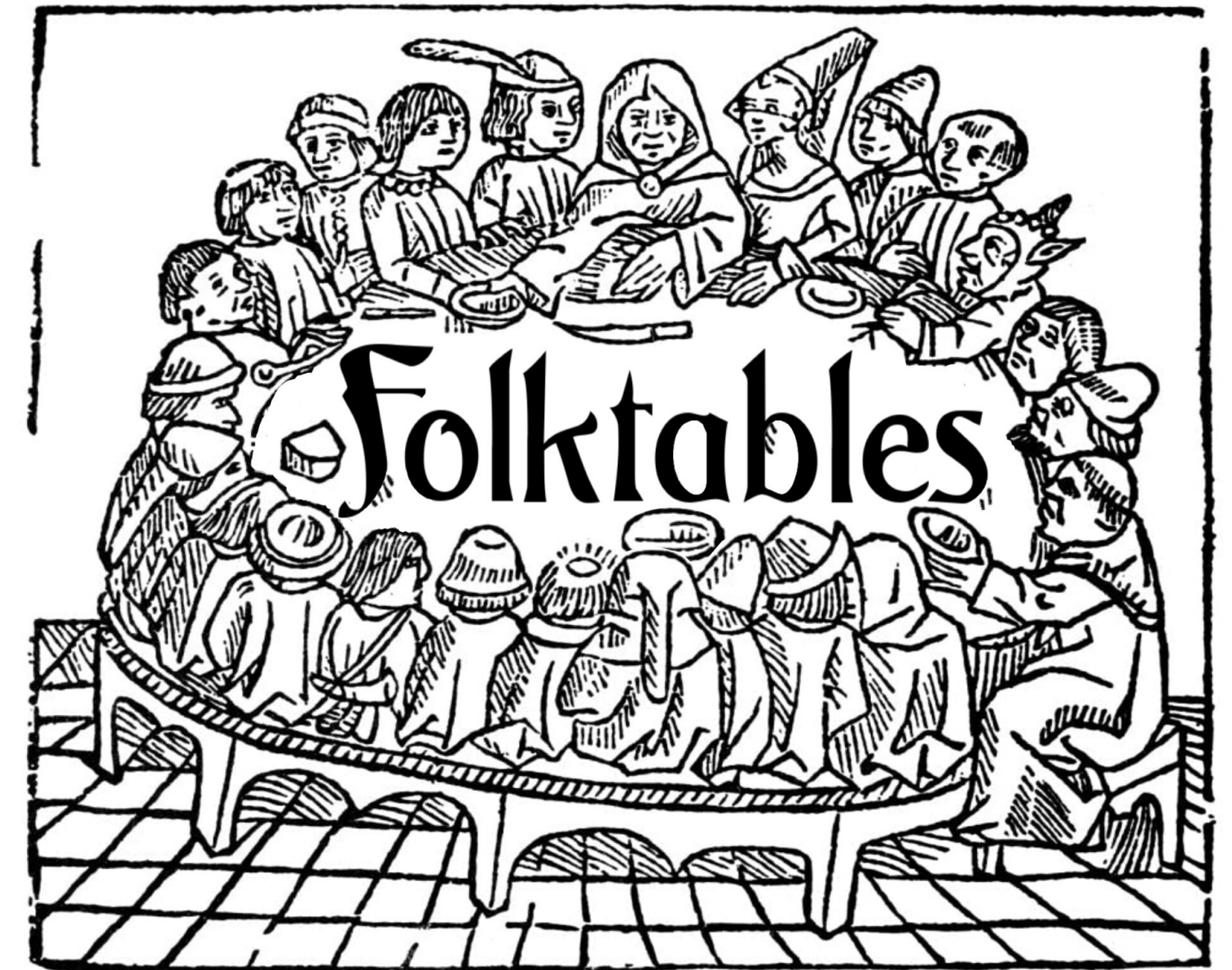
The cover of the American Community Survey questionnaire booklet. At the top left is the United States Census Bureau logo. To its right is the title "The American Community Survey". Below the title is a green-bordered box containing the text: "This booklet shows the content of the American Community Survey questionnaire." Below this is a "Start Here" section with two options: "Respond online today at: https://respond.census.gov/acs" and "Complete this form and mail it back as soon as possible." To the right of these options are input fields for "Last Name", "First Name", and "MI". Below these is a section for "Area Code + Number" with input boxes. At the bottom left, there is a section titled "Your response is required by law." with explanatory text. At the bottom right, there is a section titled "How many people are living or staying at this address?" with a bullet point: "INCLUDE everyone who is living or staying here for more than 2 months." A large, faint "ANIMATED COPY" watermark is visible across the center of the form.

Folktables: New datasets from Census microdata



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- New prediction problems from ACS data:



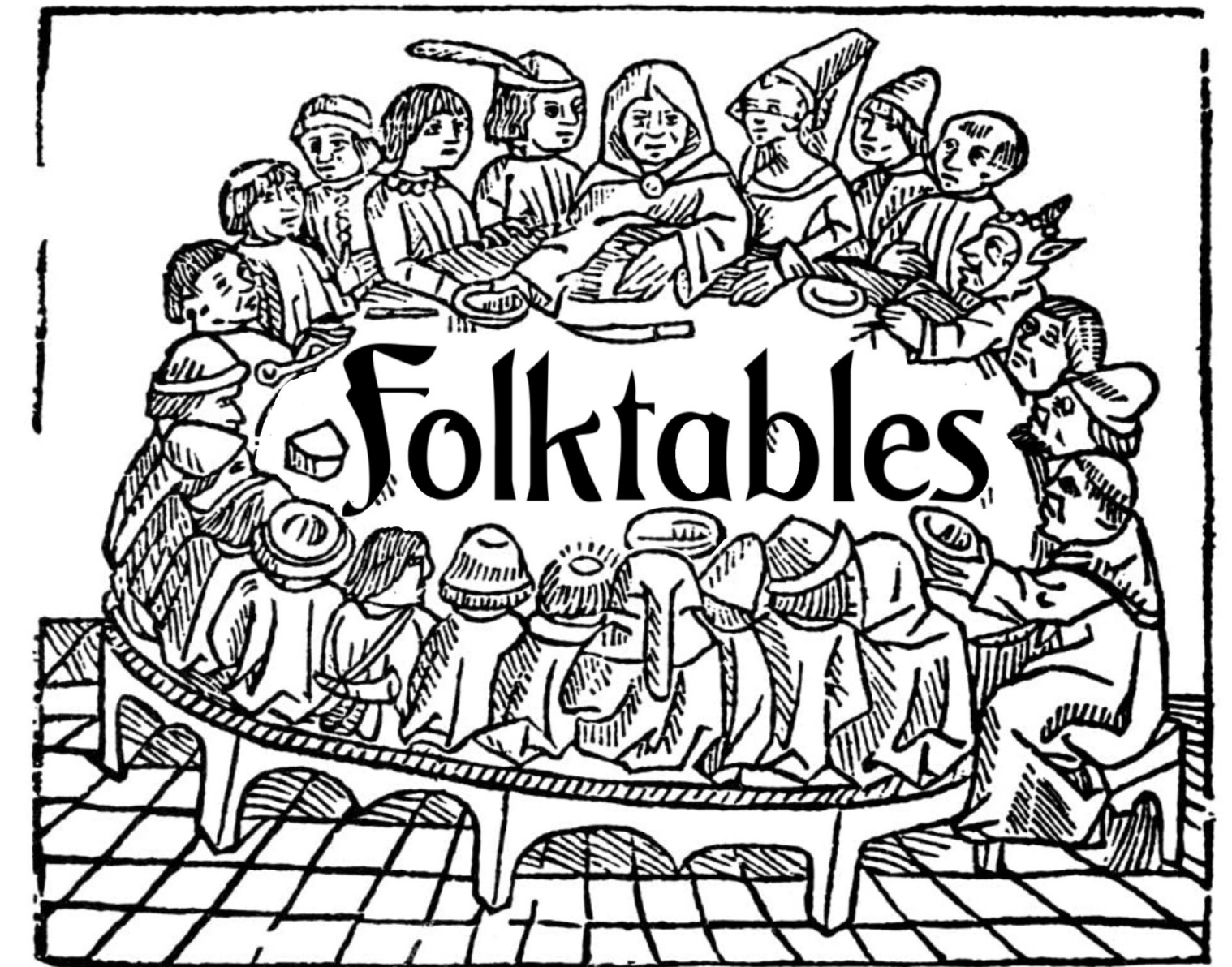
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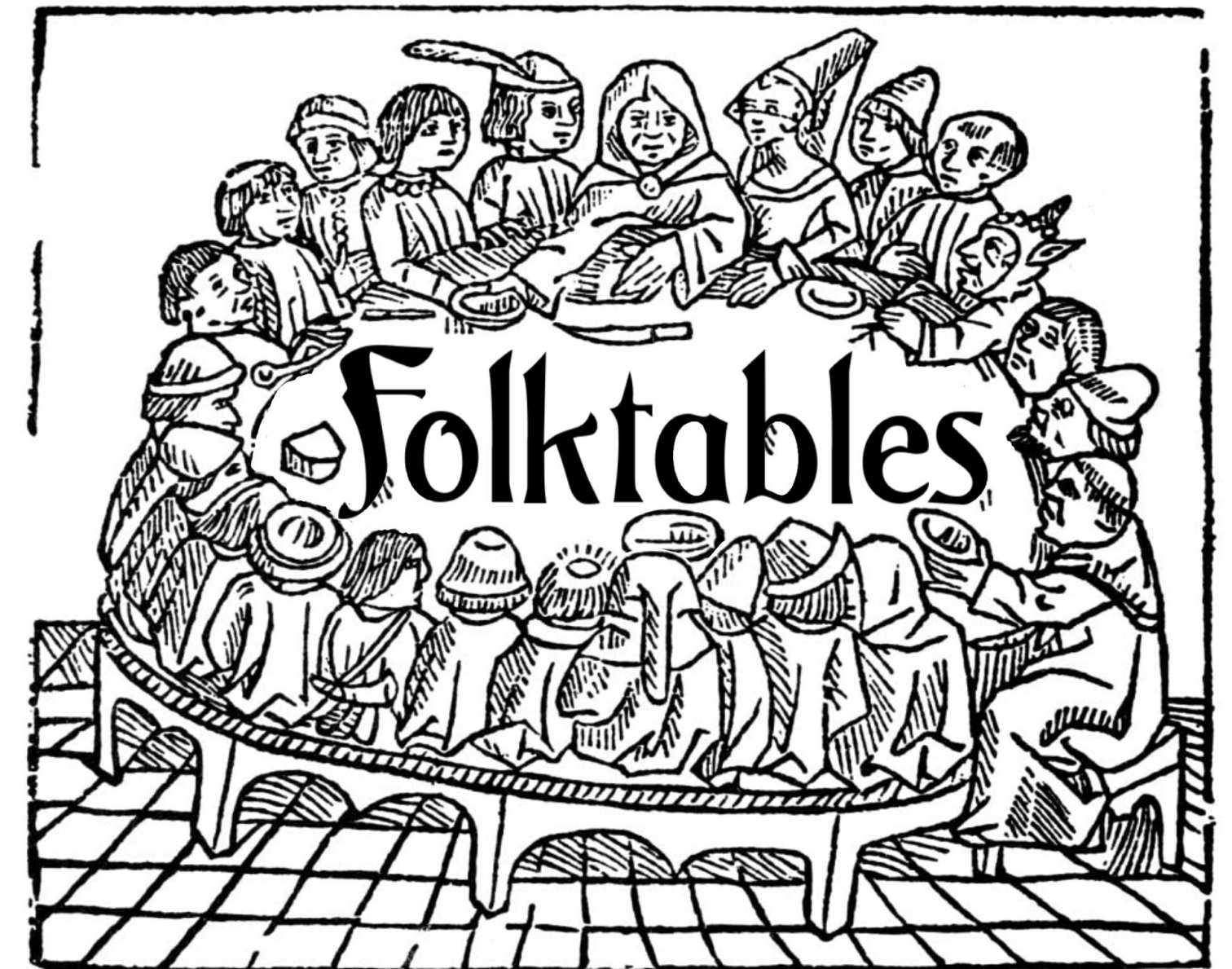
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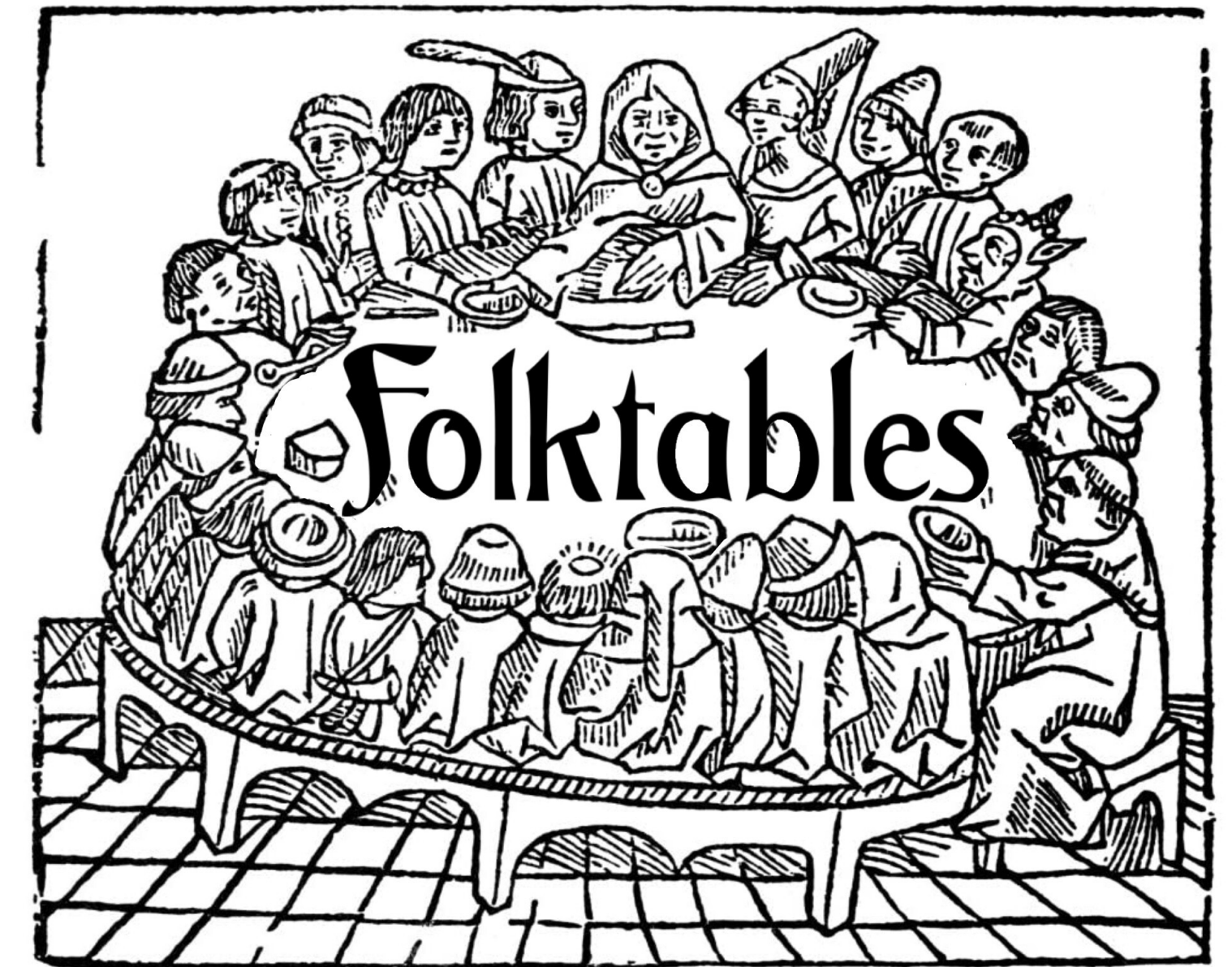
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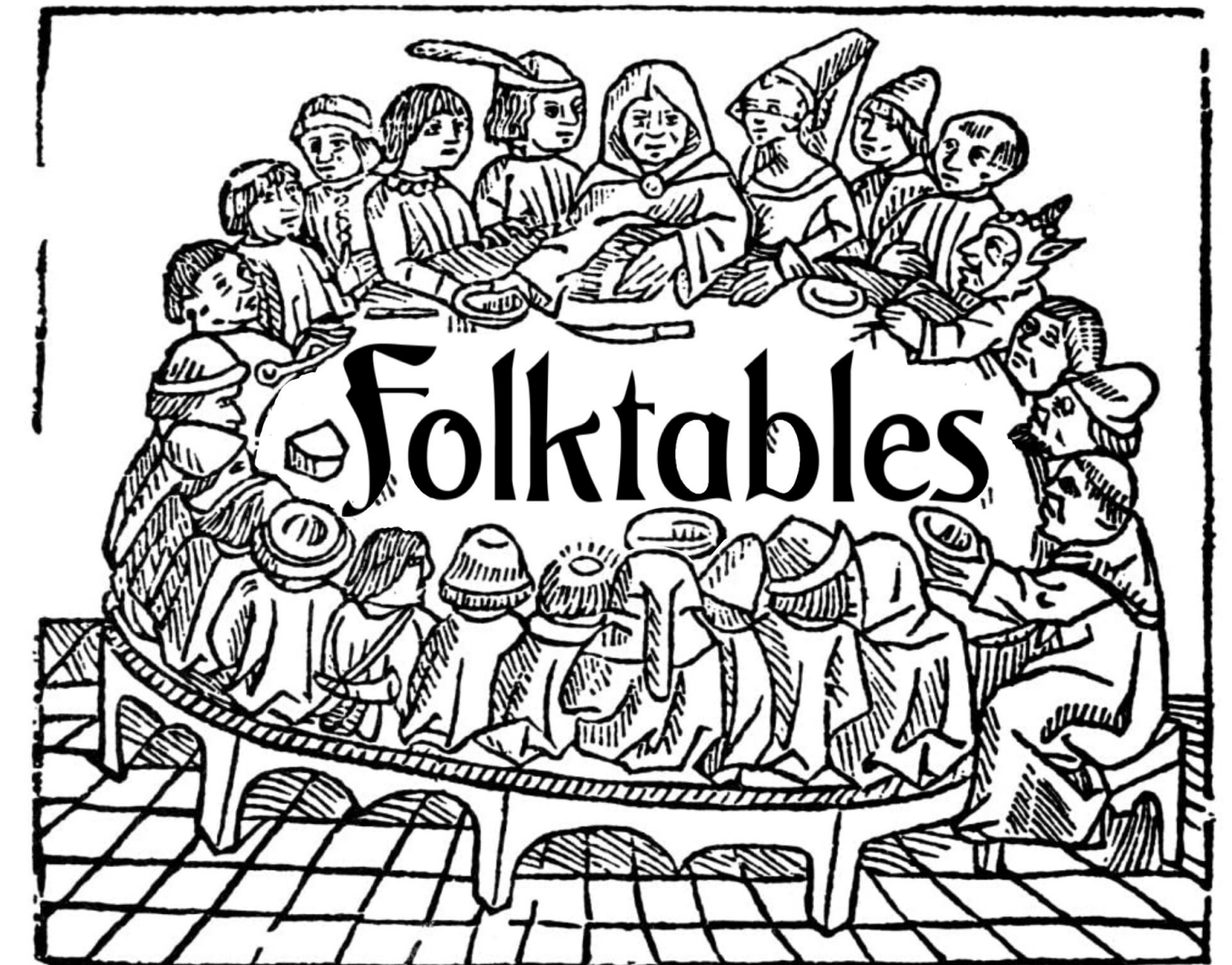
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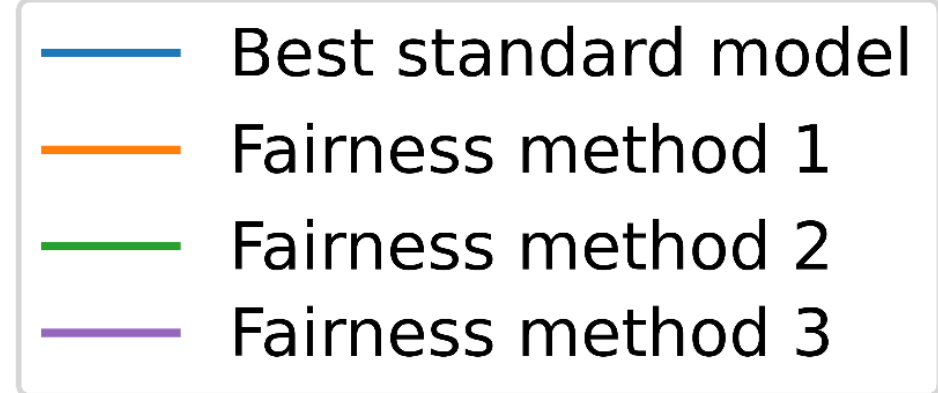
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 - Ideal for studying performance variation and distribution shift



Stability over time

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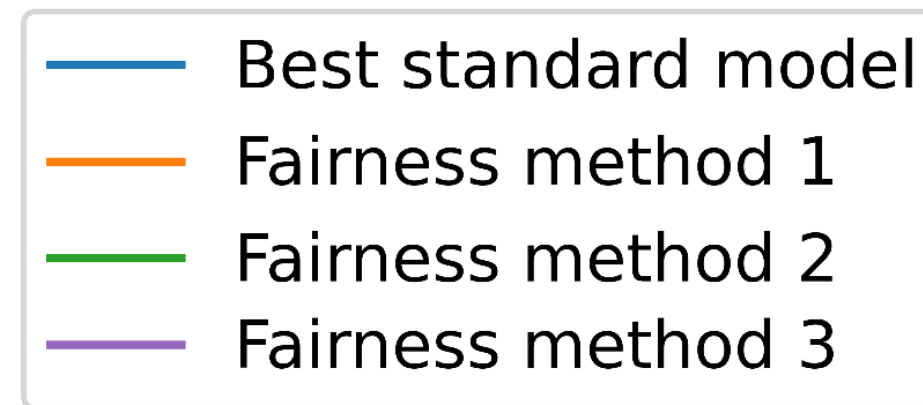
Stability over time

- Best standard model
- Fairness method 1
- Fairness method 2
- Fairness method 3

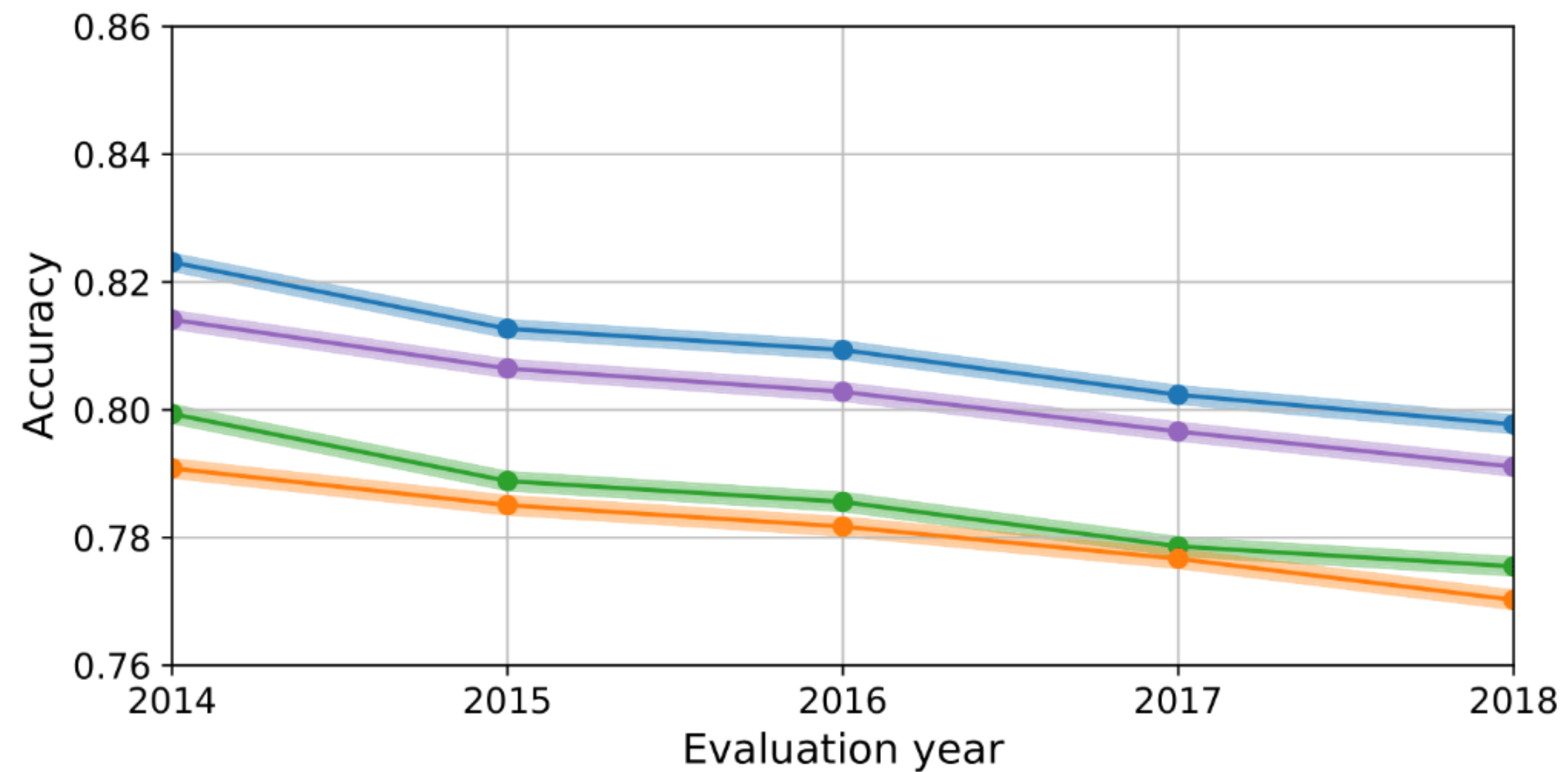
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Evaluated on 2014-2018 data

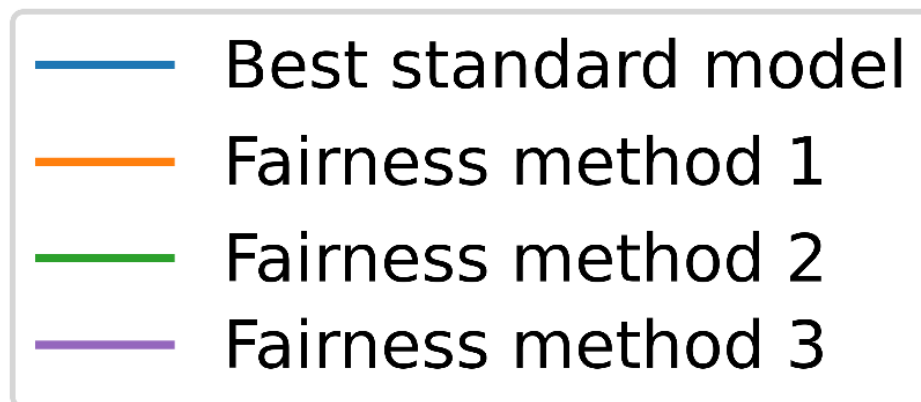
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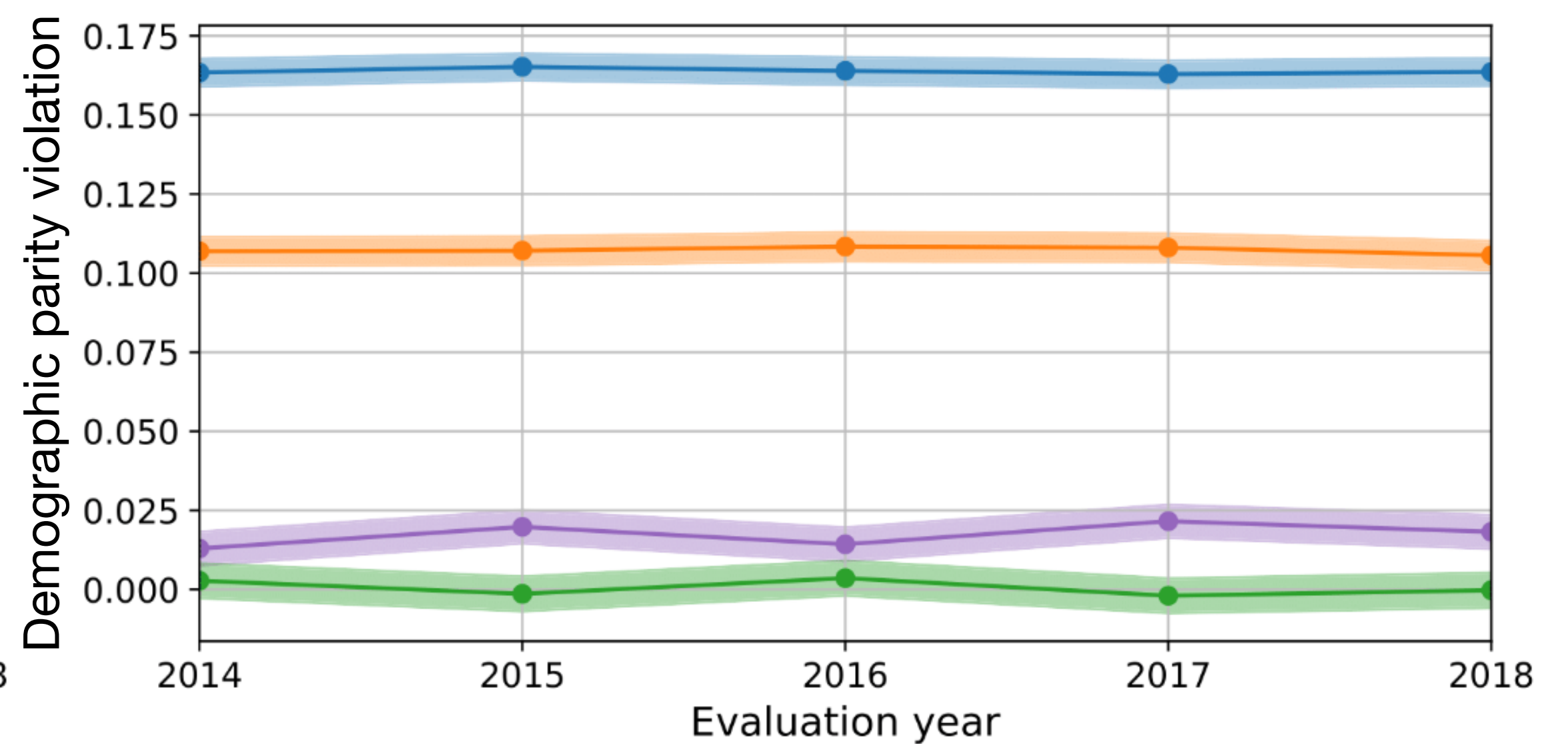
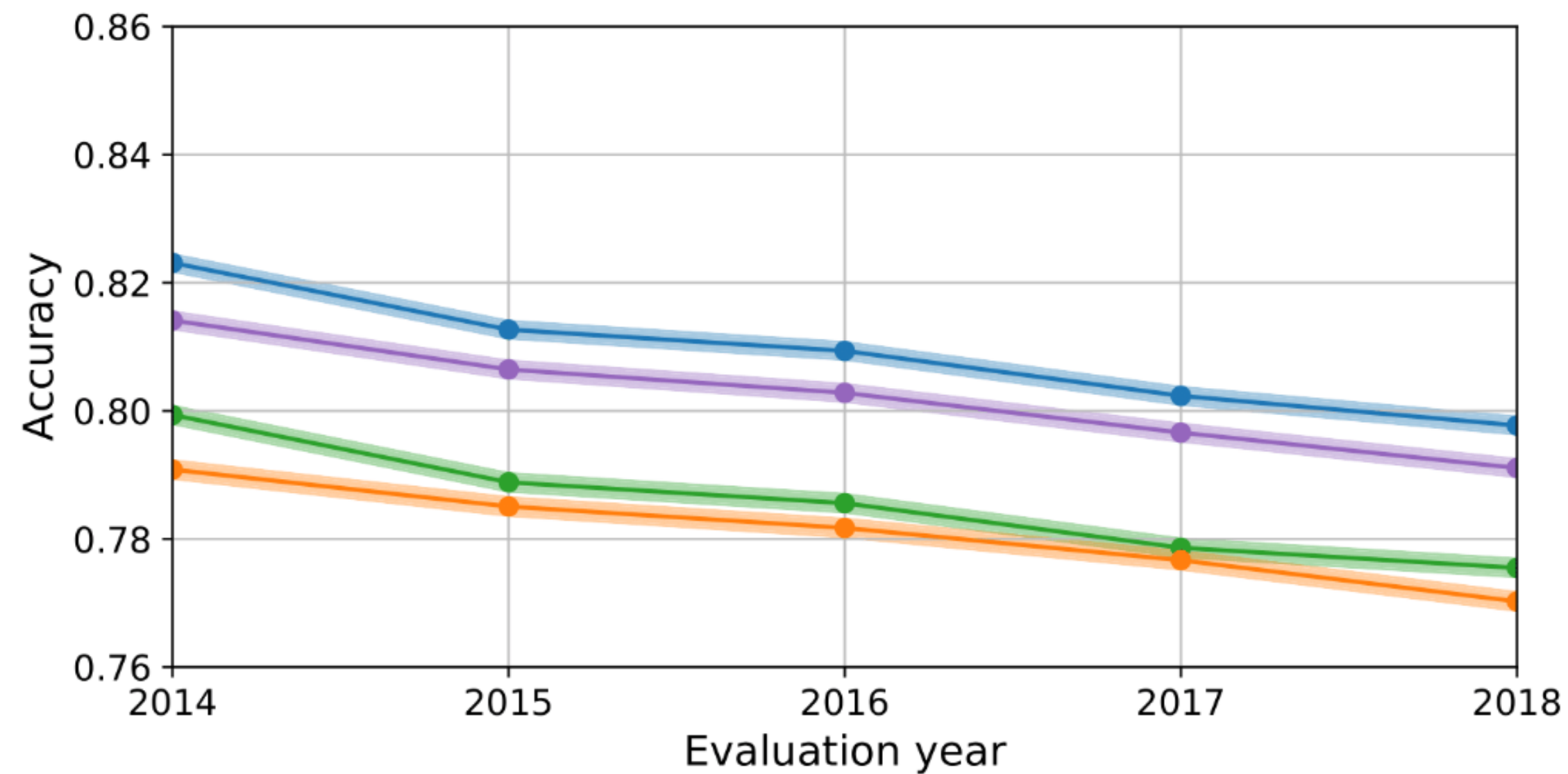
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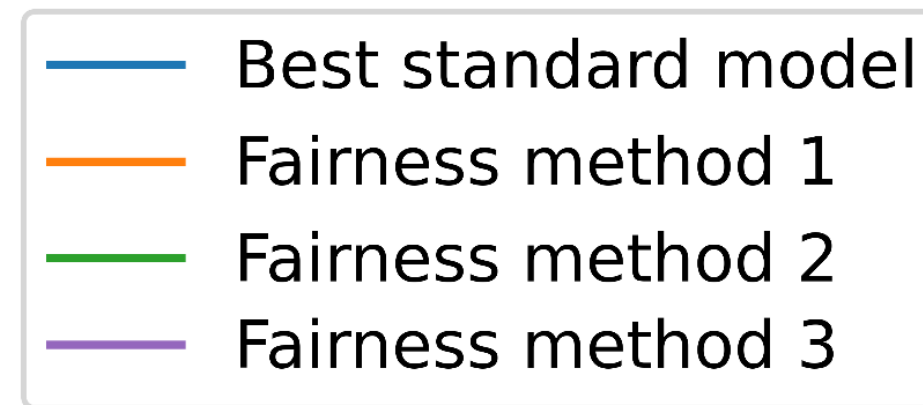
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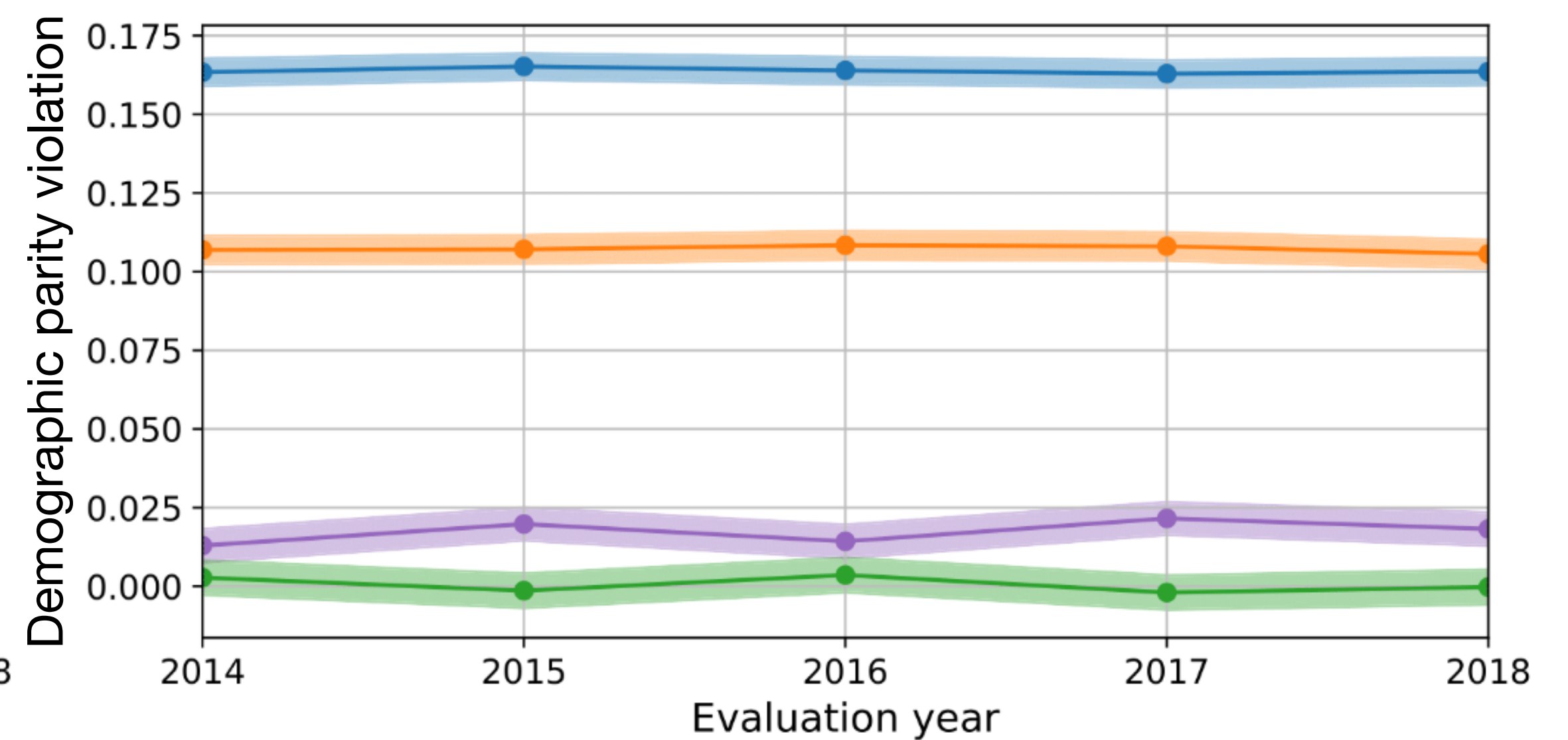
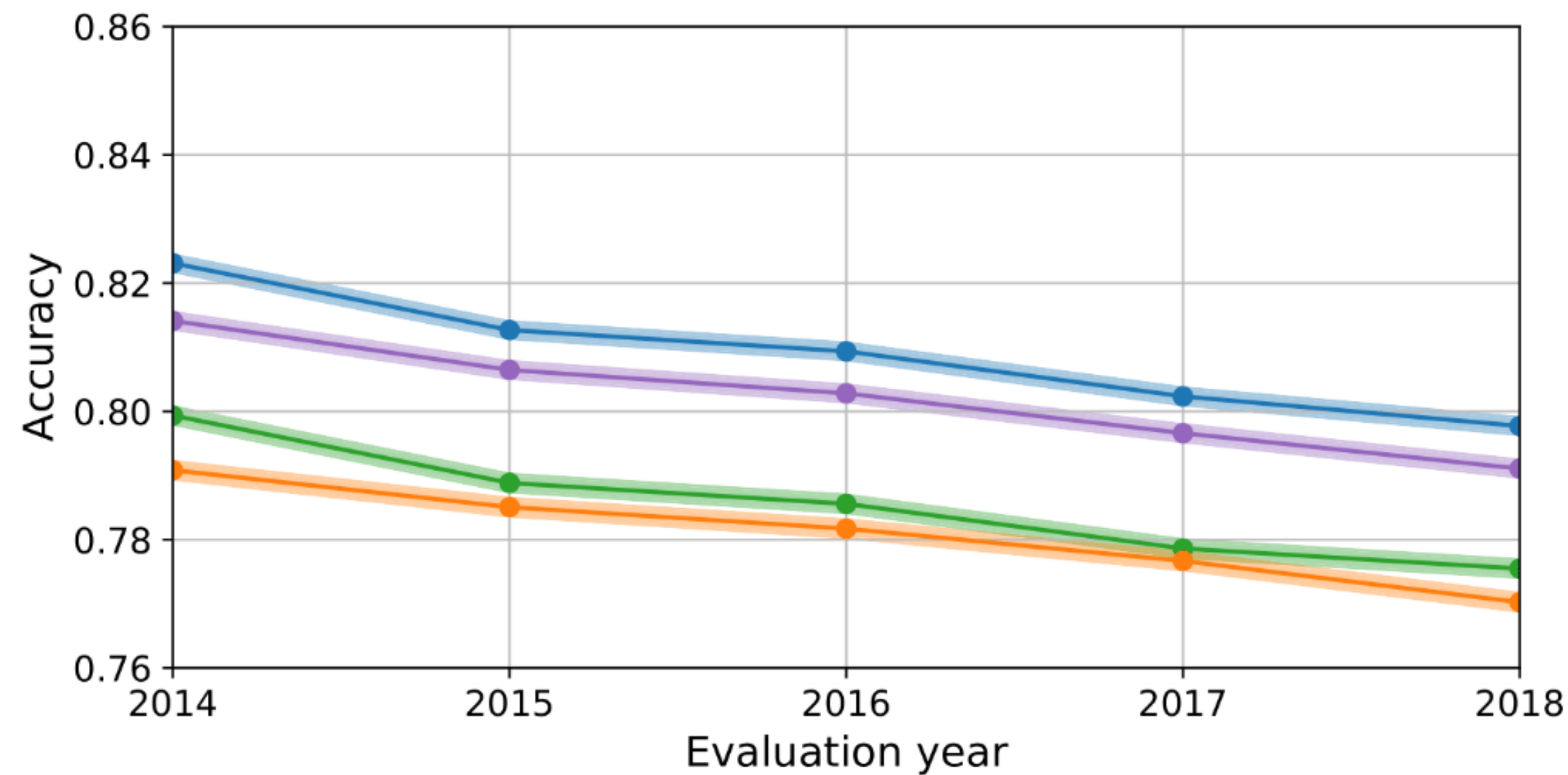
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Evaluated on 2014-2018 data



Fairness metrics are more stable over time than predictive accuracy

Transferring interventions between states

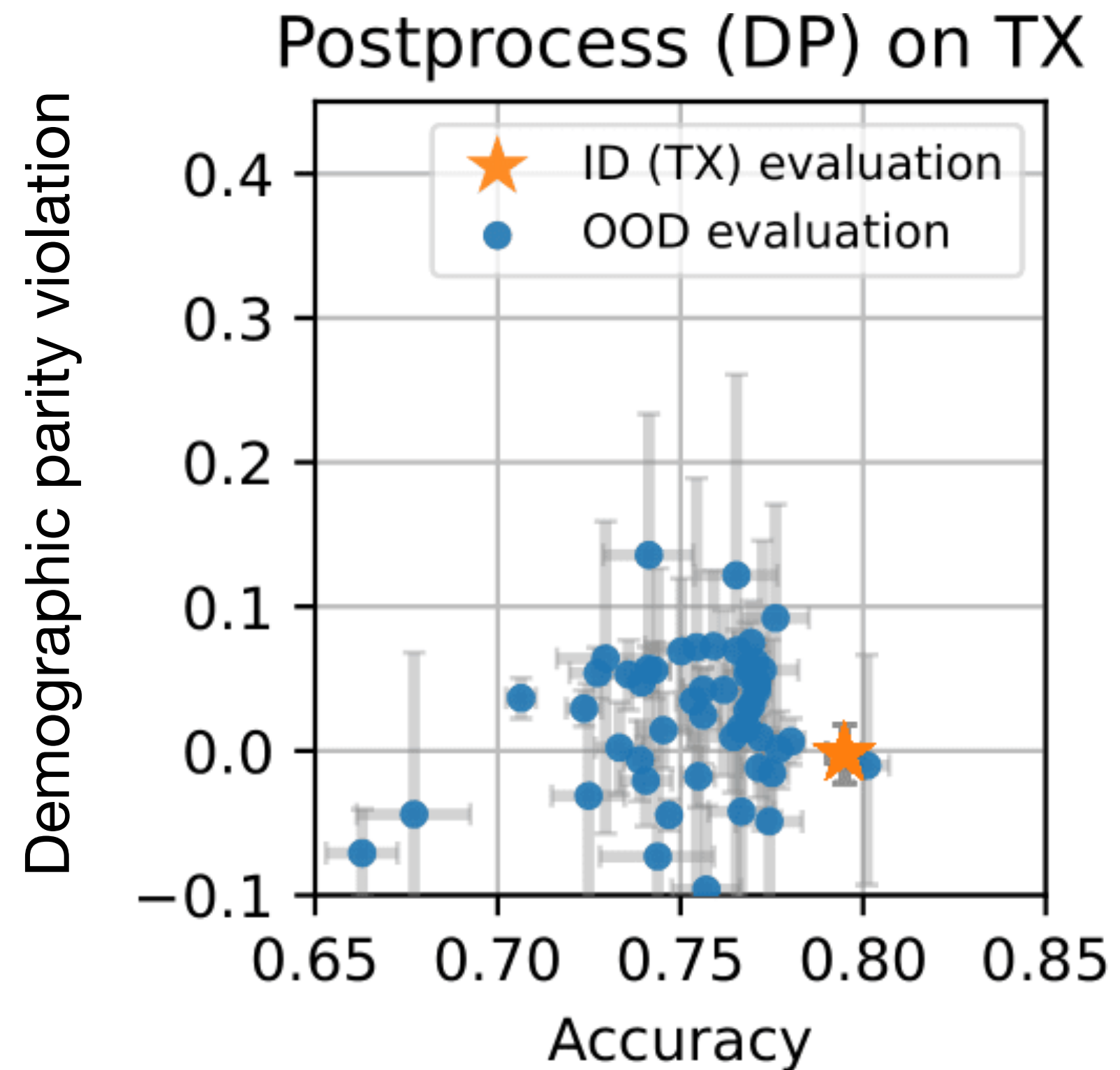
Transferring interventions between states

- Trained classifier on Texas data to enforce demographic parity

Transferring interventions between states

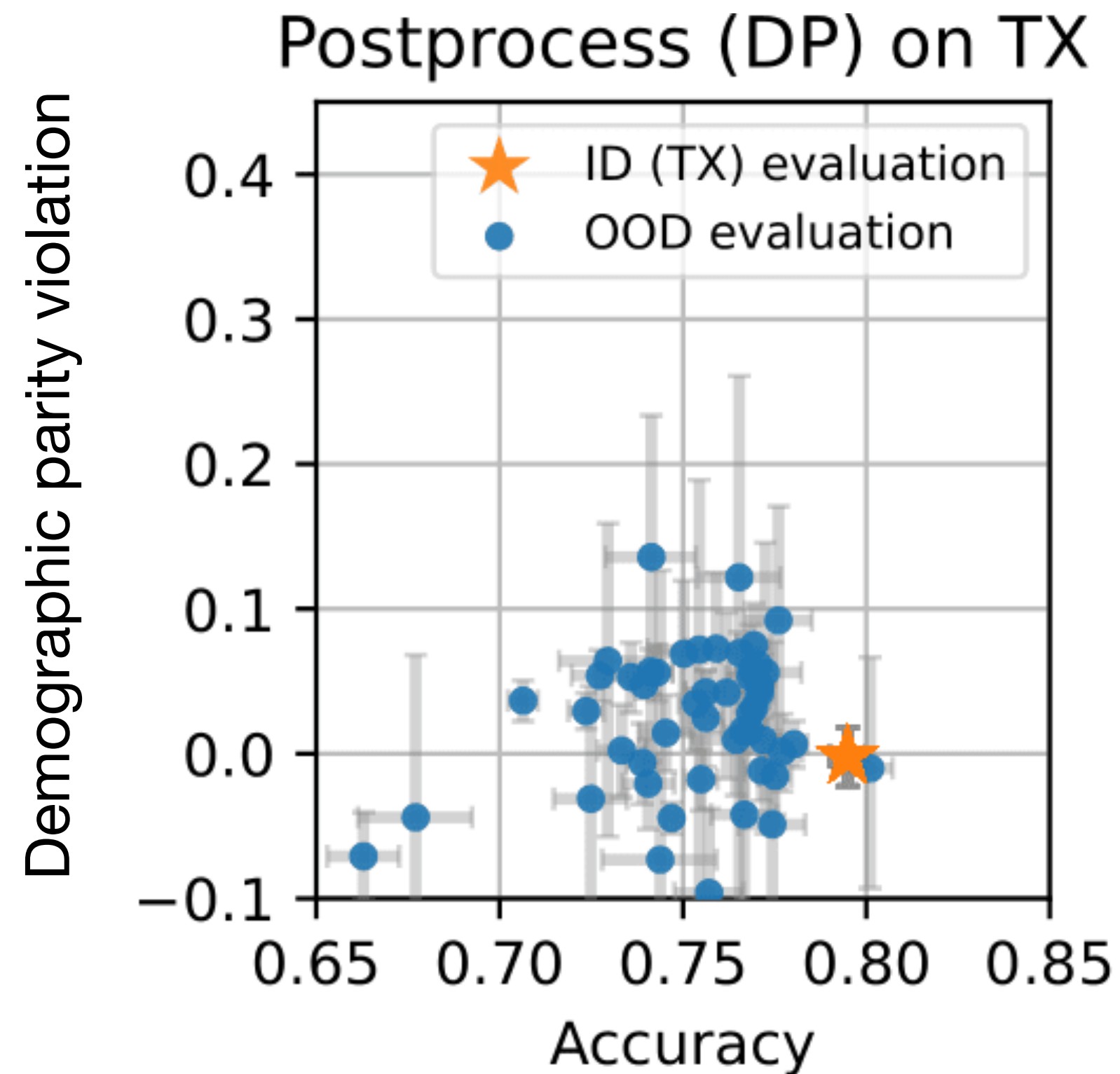
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- Evaluated classifier on all 50 states (each dot corresponds to a state)

Transferring interventions between states



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- Demographic parity violations vary substantially across states

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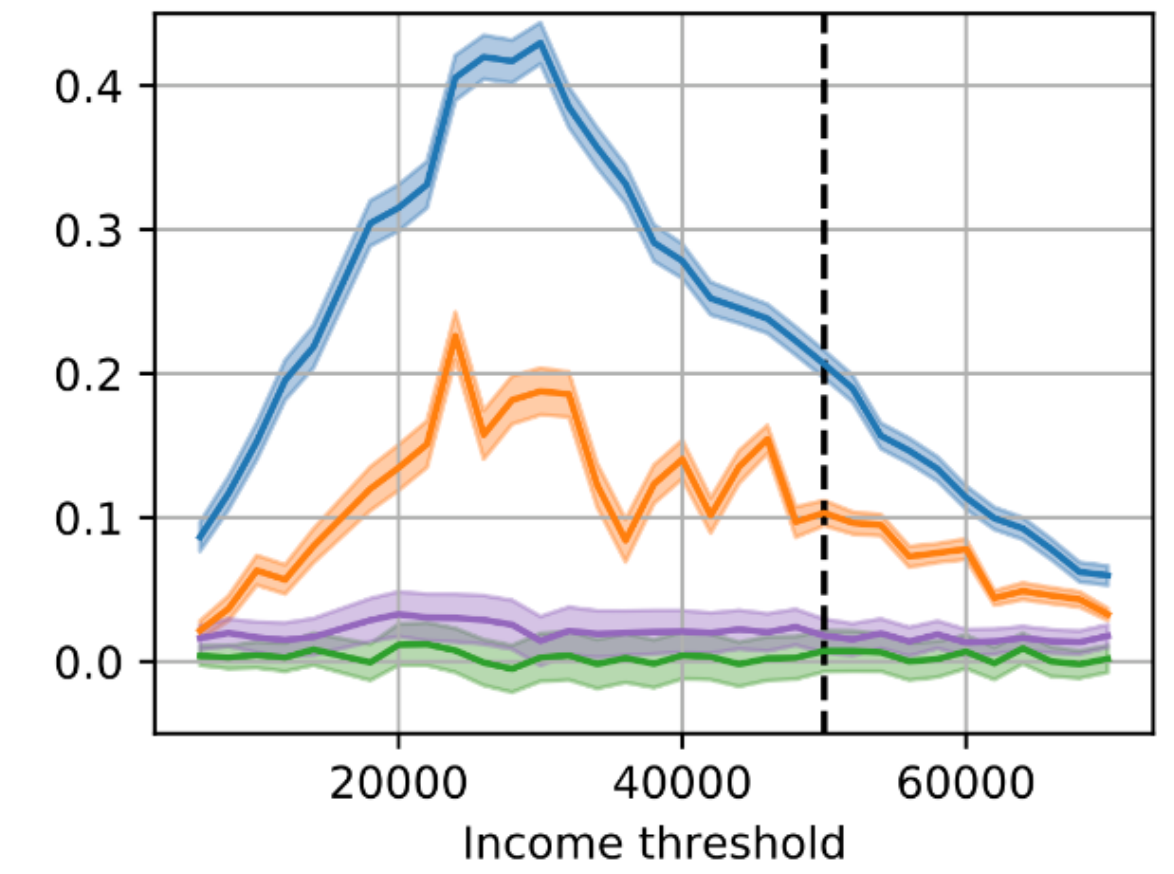
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 - **Idea:** Leverage the international census data from the 102 countries in IPUMS international

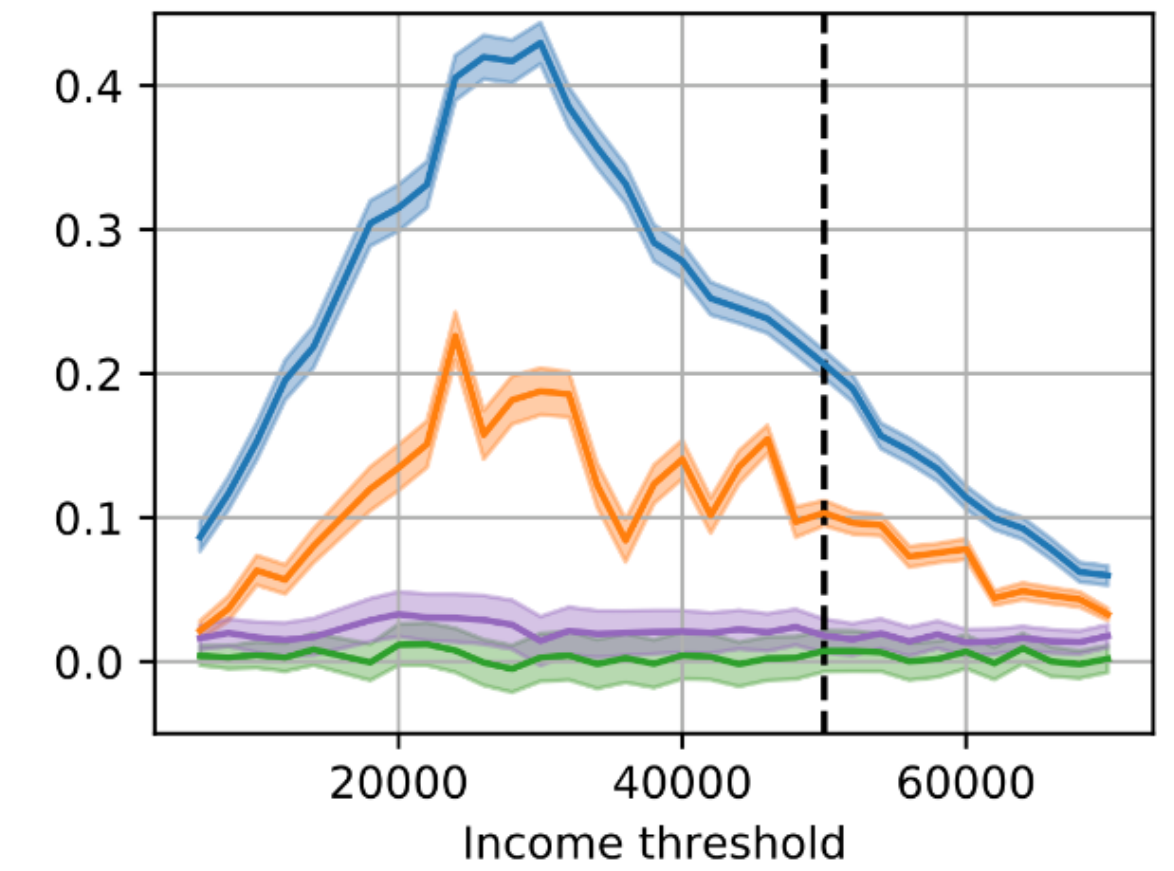


Takeaways



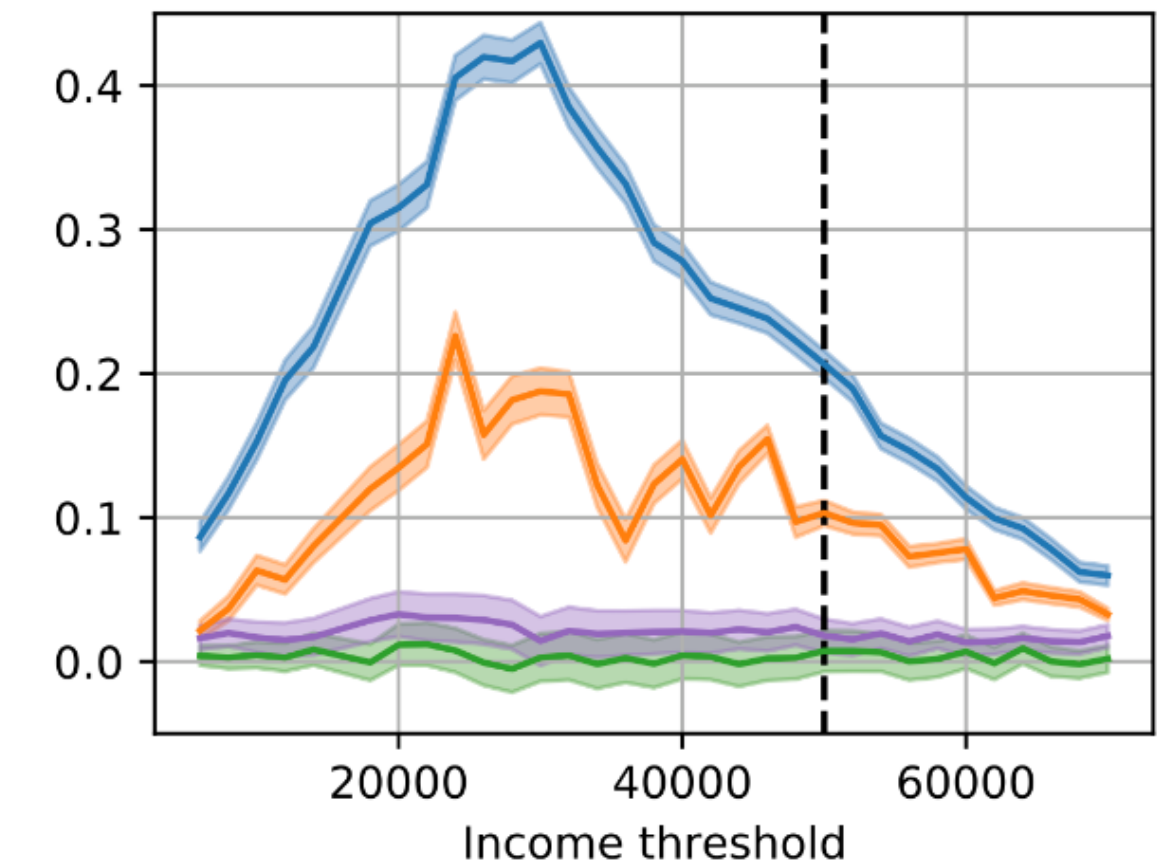
Takeaways

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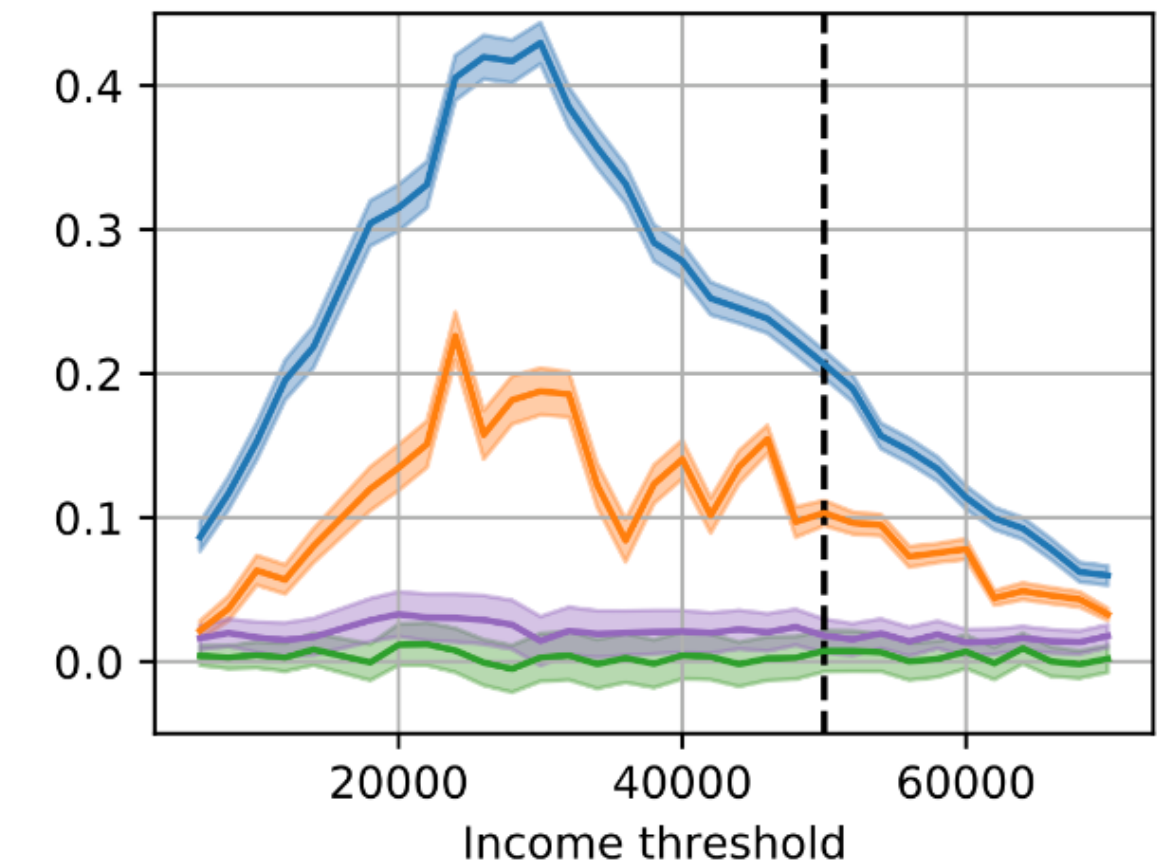
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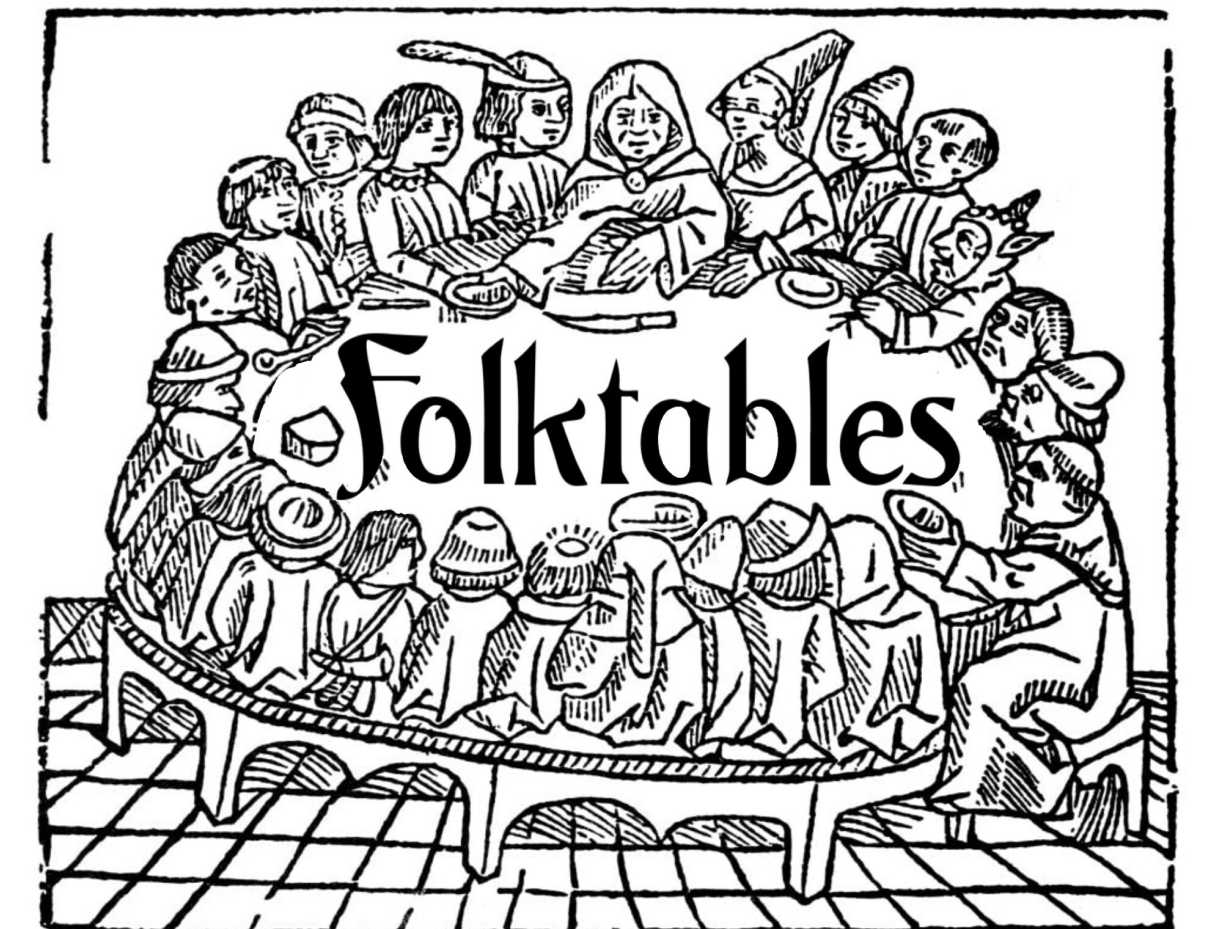
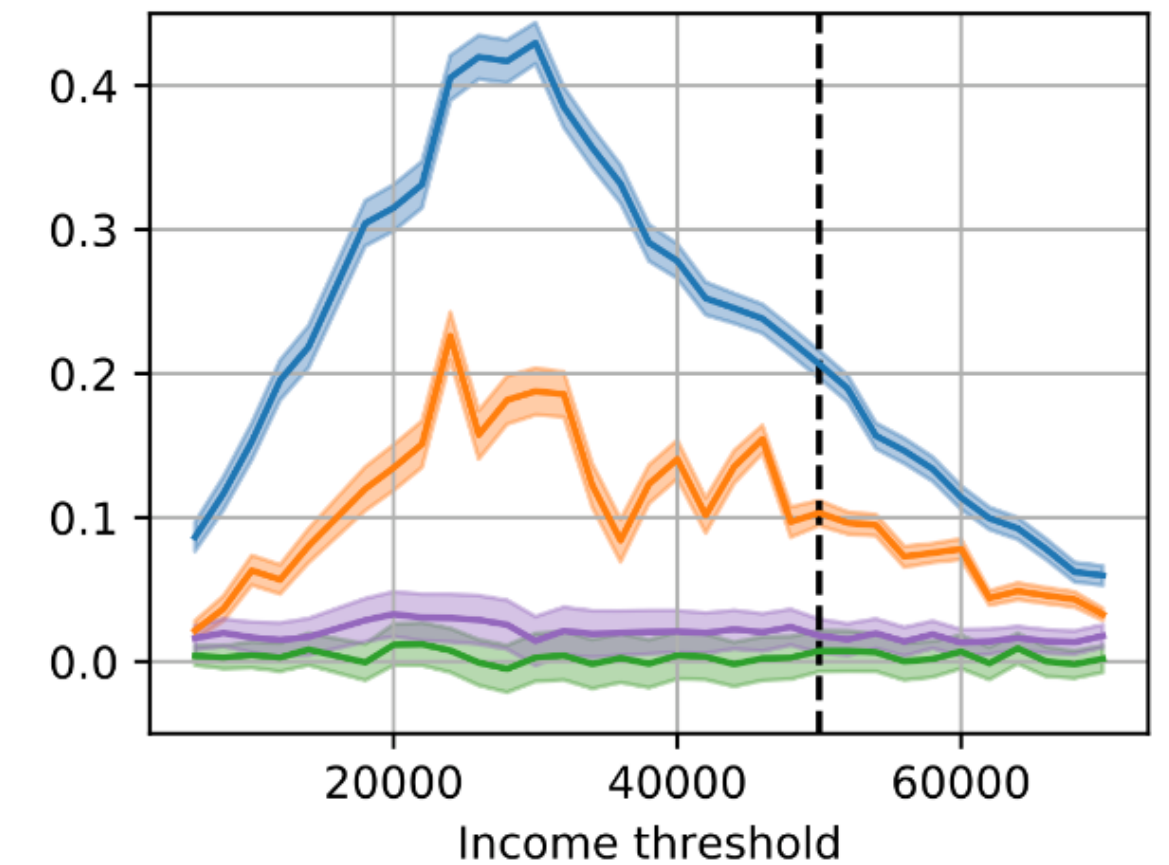
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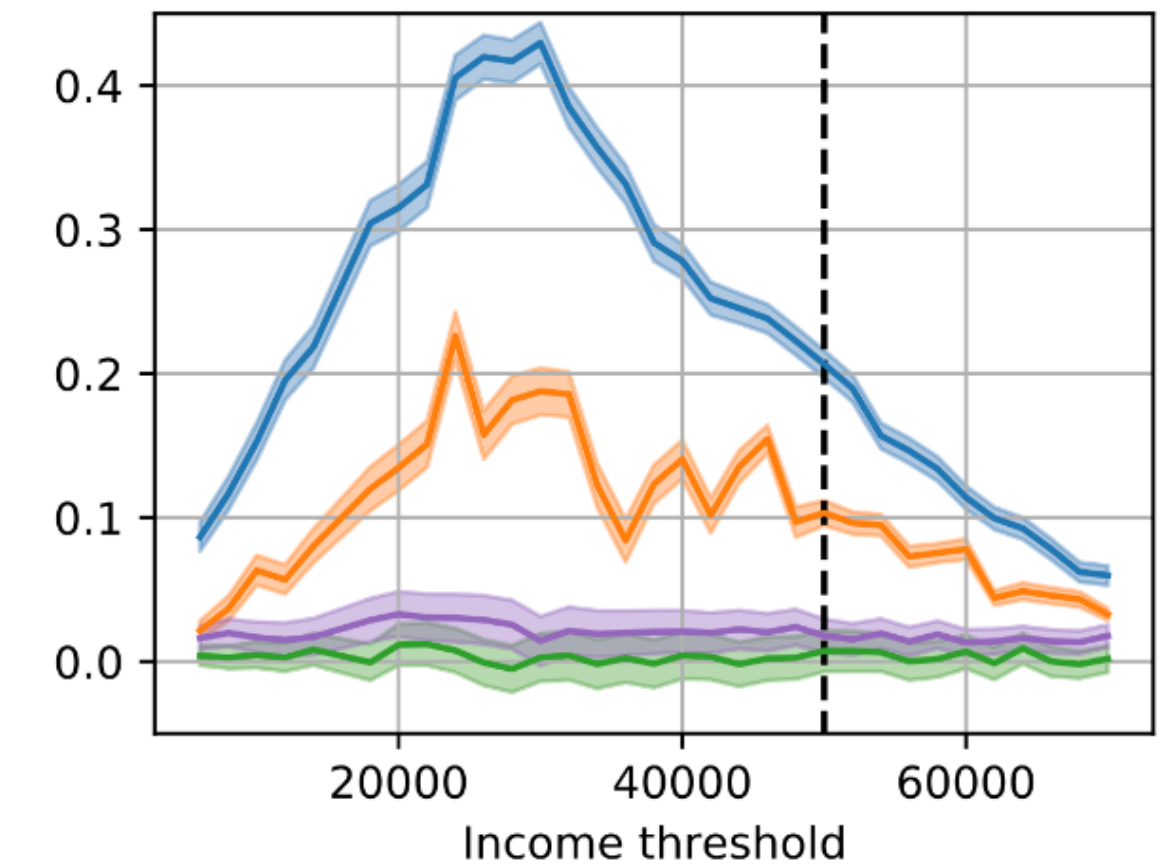
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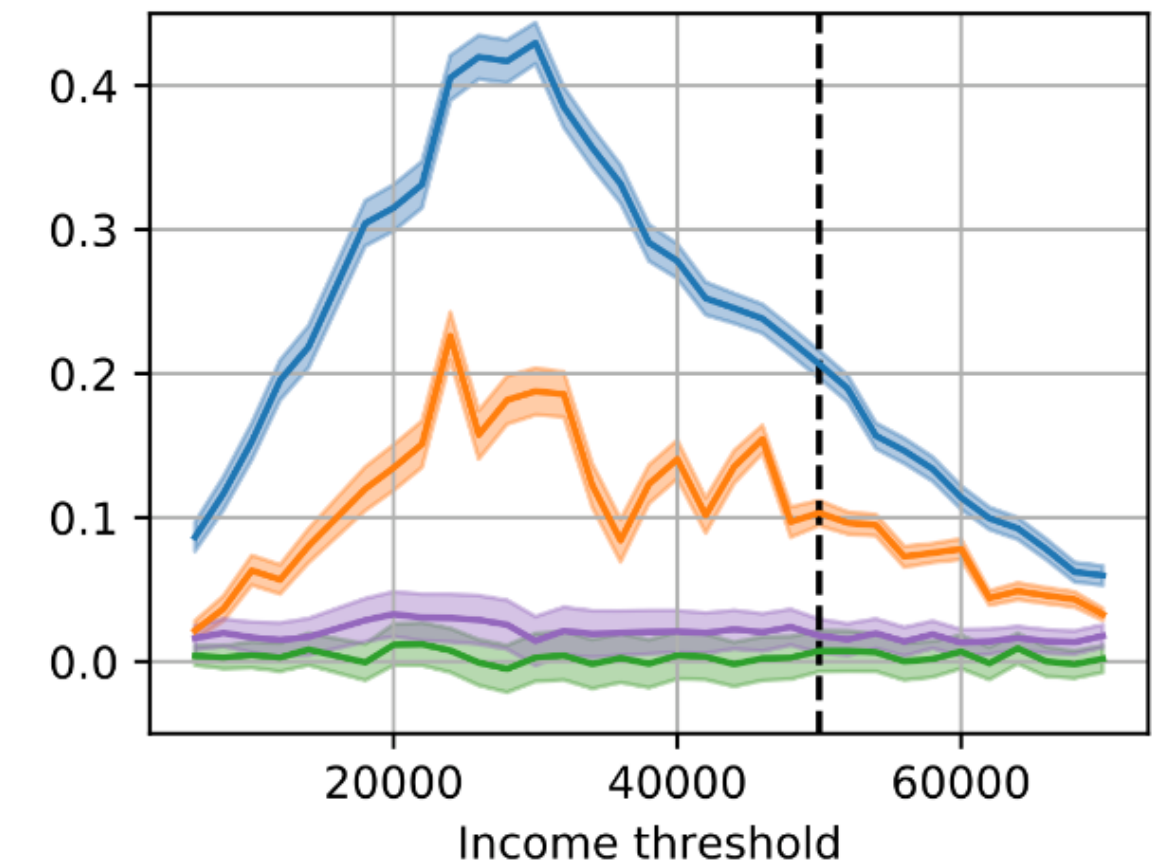
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 - Distribution shift, variations between states, etc



New datasets: <https://folktables.org>

