Explaining Racial Disparities in Personal Bankruptcy Outcomes

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Motivation

- Personal bankruptcy is a major source of debt relief for US households
 - 1 in 10 Americans have filed at some point in their life (Keys, 2018)
 - ► Average \$149k per filer ⇔ \$832/adult/year discharged annually (US Courts, 2019)
- There are significant racial disparities in financial outcomes in the US
 - Median wealth of white households is 10x Black and Hispanic wealth: (\$171k vs. \$17k) (2016 SCF)
 - Minorities pay higher interest rates than whites with the same credit score (Ghent Hernández-Murillo Owyang, 2014; Bayer Ferreira Ross, 2017, Butler Mayer Weston 2021)
 - Black household consumption falls 50% more in response to the same income shock (Ganong Jones Noel Farrell Greig Wheat, 2020)

This Paper

- **Question:** What racial disparities exist in personal bankruptcy? And why?
- Approach:
 - What observable filer characteristics explain disparities in bankruptcy outcomes?
 - Develop framework to formalize how homophily can detect and quantify racial bias
 - Estimate racial homophily between filers and judges/trustees

This Paper

- **Question:** What racial disparities exist in personal bankruptcy? And why?
- Approach:
 - What observable filer characteristics explain disparities in bankruptcy outcomes?
 - Develop framework to formalize how homophily can detect and quantify racial bias
 - Estimate racial homophily between filers and judges/trustees
- Main findings:
 - Black filers' cases are more likely to be dismissed (without debt discharge) on average
 - Chapter 7: 3 pps more often (167% higher) than non-Black filers
 - Chapter 13: 21 pps more often (41% higher) than non-Black filers
 - Observable variables reduce disparities to 0.6 and 11 pps for Chapters 7 and 13
 - Random assignment to white trustees \Rightarrow Ch 13 dismissal rate \uparrow 10 pps for Black filers

Contributions to Related Literature

- Racial disparities in household finance: Munnell, Browne, McEneaney, and Tootel (1996); Braucher et al. (2012); Reid Bocian, Li, and Quercia (2017); Bayer et al. (2018); Bartlett, Morse, Wallace, and Stanton (2019); Fuster et al. (2020); Morse and Pence (2020); Blattner and Nelson (2021); Begley and Purnanandam (2021); Dobbie Liberman Paravisini (2021); Goldsmith-Pinkham, Scott, and Wang (2022)
 - New focus on racial disparities in **bankruptcy** and its drivers
- Impact of legal decision-makers: Anwar et al. (2012, 2019a, 2019b); Arnold, Dobbie, and Yang (2018); Morrison et al. (2019); Arnold, Dobbie, and Hull (2020); Iverson (2020); Iverson et al. (2020)
 - Evidence on role of bias and importance of bankruptcy trustees
- **Methods for detecting and quantifying bias:** Becker (1957, 1993); Knowles, Persico, and Todd (2001); Anwar and Fang (2006); Arnold, Dobbie, and Yang (2018); Arnold, Dobbie, and Hull (2020); Canay, Mogstad, and Mountjoy (2020); Hull (2021); Bohren, Hull, and Imas (2022)
 - New results formalize how can homophily detect and quantify bias
 - ► Homophily can detect bias in **cases where outcome tests are infeasible**

Background and Data

What is Personal Bankruptcy?

- Discharge unsec. debt (credit card, medical, etc.); make partial payments to creditors
- Households file under one of two Chapters:
 - Chapter 7: discharge received upon initial legal ruling (~3 month process)
 - **Chapter 13:** discharge received after completing 5 year repayment plan
- Three important legal decision makers (DMs):
 - Judge: ultimately decides case outcomes (e.g., dimissal)
 - Trustee: evaluates filer's accuracy and honesty; makes recommendations to judge
 - Attorney: advises filer on Chapter choice and reporting

Bankruptcy Outcomes

- Possible case outcomes: discharge, conversion of chapter, and dismissal
- What are the main reasons for **dismissal**?
 - Fraudulent reporting by filer (e.g., concealing property)
 - ► Failure to make promised payments in Chapter 13 over 5-year period
- **Trustees** and **judges** make **subjective** evaluations of filers
 - Procedural error vs. intentional fraud?
 - Did Chapter 13 payments stop due to severe hardship beyond filer's control?
 - Assessment of feasibility of filer's Chapter 13 repayment plan

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- Trustees and judges make subjective evaluations of filers
 - Procedural error vs. intentional fraud?
 - Did Chapter 13 payments stop due to severe hardship beyond filer's control?
 - Assessment of feasibility of filer's Chapter 13 repayment plan
 - Outcomes test isn't feasible when outcome(s) DM values are unobserved by researcher

Bankruptcy Data

- Lexis Nexis bankruptcy case data
 - Filer names and addresses, chapter, events during case, case outcomes, and DM names
 - Near universe of US bankruptcy cases: > 63 million cases
 - ► Full coverage of US Jan. 2010 Jun. 2022
- Federal Judicial Center (FJC) case data
 - Additional case info for 2008+
 - Includes filer assets, liabilities, and income
- 2021 FL Voter Registration Records: 20M obs, used to predict filer and DM race
 - ▶ Used to train and test deep-learning race-imputation model (based on Kotova, 2022)
 - ▶ Use full names and addresses (tract-level race composition); achieve 85% accuracy

Model Perf. Stats >> Model Perf. Graphs

Racial Disparities in Bankruptcy Dismissals

Racial Disparities in Dismissal Rates



Obs: 39M–12M (Ch. 7), 14M–4M (Ch. 13); **Clustering:** ZIP (95% confidence interval shown); **Case controls:** 1[pro se], 1[prior filing], 1[nonexempt assets], 1[homeowners], 1[joint filing], ln(assets), debt/assets, % secured debt, ln(income), and income - expense gap

Homophily and Bias: Decision & Econometric Model

Decision Model: Setting and Notation

- A DM *j* with race $r_j \in \{b, w\}$ observes filer *i*'s race $r_i \in \{b, w\}$ and non-race char. *x*
- The DM chooses whether to dismiss $D \in \{0, 1\}$ to maximize her expected utility
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- The DM solves: $\max_{d \in \{0,1\}} E_j[u(Y_d; j, r_i, x) | r_i, x]$
 - The DM takes expectation wrt to her beliefs
- Denote her payoff from dismissing: $\Delta(j, r_i, x) \equiv u(Y_1; j, r_i, x) u(Y_0; j, r_i, x)$
- Her optimal decision is $D(j, r_i, x) = 1\{E_j[\Delta(j, r_i, x)|r_i, x] \ge 0\}$

Sources of Bias/Discrimination

- Denote the DM's prediction error: $\mu(j, r_i, x) \equiv E[\Delta(j, r_i, x)|r_i, x] E_j[\Delta(j, r_i, x)|r_i, x]$
 - Differential prediction error by filer race can lead to inaccurate statistical discrimination
- Denote taste for discrimination: $\beta(j, r_i, x) \equiv E[\Delta(j, w, x)|r_i, x] E[\Delta(j, b, x)|r_i, x]$

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- We can decompose the DM's payoff:

$$E_{j}[\Delta(j, r_{i}, x)|r_{i}, x] = \underbrace{E[\Delta(j, w, x)|r_{i}, x]}_{\text{acc. stat. disc.}} - \underbrace{\mu(j, r_{i}, x)}_{\text{inacc. stat. disc.}} - \underbrace{1[r_{i} = b]\beta(j, b, x)}_{\text{taste-based disc.}}$$

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• We can similarly decompose the drivers of the DM's decision:

$$D(j, r_i, x) = \underbrace{\widetilde{D}(j, r_i, x)}_{\text{decision w/ only acc. stat. disc.}} + \underbrace{\widetilde{\beta\mu}(j, r_i, x)}_{\text{influence of }\mu \text{ and }\beta}$$

where $\widetilde{D}(j, r_i, x) \equiv 1{E[\Delta(j, w, x)|r_i, x] \ge 0}$ and $\widetilde{\beta\mu}(j, r_i, x) \equiv D(j, r_i, x) - \widetilde{D}(j, r_i, x)$

Types of Bias

• A case exhibits total racial bias against Black filers if D(j, b, x) > D(j, w, x)

• A case exhibits $\beta \mu$ -racial bias against Black filers if $\widetilde{\beta \mu}(j, b, x) > \widetilde{\beta \mu}(j, w, x)$

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- Motivation for focus on βμ-racial bias:
 - Changing dismissal decisions to reduce taste-based and inaccurate statistical discrimination increases average DM welfare (net of tastes for discrimination)
 - Changing decisions due purely to accurate statistical disc. reduces average DM utility

Causal Parameters of Interest

• Average total racial bias: $\delta^{ATT} \equiv E[D(j, b, x) - D(j, w, x)|r_i = b]$

• Average
$$\beta\mu$$
-racial bias: $\delta^{\beta\mu} \equiv E[\widetilde{\beta\mu}(j, b, x) - \widetilde{\beta\mu}(j, w, x)|r_i = b]$

• Identification challenges:

1 Average difference in dismissal rates could reflect selection (x correlated with r_i)

2 Isolating $\beta\mu$ -racial bias from total racial bias

Homophily Estimand & Parallel Disparities Assumption

• Homophily estimand: $\tau \equiv \{E_{bw}[D] - E_{ww}[D]\} - \{E_{bb}[D] - E_{wb}[D]\}$

racial disparity w/i white trustees racial disparity w/i Black trustees

► To minimize notation, we write: $E[D|r_i, r_j] = E_{r_i r_j}[D]$ and suppress D's dependencies

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• Assumption 1 (Parallel Disparities):

$$E_{bw}[D(w)] - E_{ww}[D(w)] = E_{bb}[D(w)] - E_{wb}[D(w)]$$

I.e., the difference in Black/white filer outcomes due to non-race characteristics, which may be correlated with race, is the same among filers assigned to white or Black DMs

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- Could fail if conditional distribution of *x*|*r_i* varies with DM race (in practice, this could arise if DM and filers could choose to work together)
 - ► Falsification test: do *r_i* and case observables predict *r_j*?
- Could fail if Black/white DM decisions respond diff. to non-race chars corr. w/ race
 - ► Falsification test: does relationship between *D* and case chars vary with DM race *r_i*?

Identifying Total Racial Bias

 Prop 1: IFF Assumption 1 (Parallel Disparities) holds, the homophily estimand identifies the average difference in total racial bias between Black and white DMs:

$$\tau = \delta_W^{ATT} - \delta_B^{ATT}$$

- \Rightarrow testing H_0 : $\tau = 0$ serves as test for the presence of total racial bias
 - If $\tau \neq 0$, at least one DM exhibited total bias
 - > Note: test has exact size, but may fail to reject when total bias is present

Parallel Accurate Statistical Discrimination

Assumption 2 (Parallel Accurate Statistical Discrimination, AKA PASD):

$$E_{bw}[\widetilde{D}(b) - \widetilde{D}(w)] = E_{bb}[\widetilde{D}(b) - \widetilde{D}(w)]$$

I.e., if DMs make decisions based only on accurate statistical discrimination, the effect of a Black filers' race on dismissal would be equal across both white and Black DMs

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 - Hence same tests can help falsify this assumption

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- Faces similar possible violations as parallel disparities
 - Hence same tests can help falsify this assumption
- **Prop 2:** Under Assumption 1 (Parallel Disparities), IFF Assumption 2 (PASD) holds, the homophily estimand identifies the avg. diff. in βμ-racial bias btwn Black/white DMs:

$$\tau = \delta_W^{\beta\,\mu} - \delta_B^{\beta\,\mu}$$

• \Rightarrow testing H_0 : $\tau = 0$ serves as test for the presence of $\beta \mu$ -racial bias

Quantifying Bias

- Under parallel disparities and PASD , homophily estimand captures relative bias
- Assumption 4: $\delta_B^{\beta \mu} \ge 0$ (on avg., Black DMs weakly exhibit bias against Black filers)
- Is Assumption 4 plausible?
 - Psychology research documents pro-white implicit bias among US minorities (Nosek et al., 2002; Livingston, 2002; Ashburn-Nardo et al., 2005)
 - Black patients exhibit higher WTP for white doctors vs. Black doctors (Chan, 2022)

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- Under Assumptions 1-2 and Assumption 4, homophily partially identifies disparity due to $\beta\mu$ -racial bias:

$$\delta^{\beta\,\mu} \in [(1-p) au, 1-p au]$$

where $1 - p = Pr(r_j = w)$, i.e., the proportion of white DMs

> Paper details (weaker) lower bounds obtained under weaker assumptions

Racial Homophily in Bankruptcy

Estimating DM Homophily

- To investigate the scope for bias, we examine homophily
 - ► Today, we focus on Black-white homophily between filers and trustees

• Using case-level data, we estimate

$$\begin{split} 1[\textit{Dismissed}_i] &= \beta_1 \textit{Pr}(\textit{BlackFiler}_i) + \beta_2 \textit{Pr}(\textit{WhiteTrustee}_i) \\ &+ \beta_3 \left[\textit{Pr}(\textit{BlackFiler}_i) \times \textit{Pr}(\textit{WhiteTrustee}_i)\right] + X_i \gamma + \varepsilon_i \end{split}$$

• Fixed effects: disposition year, district, filer ZIP, judge, and trustee

Identification: Random and Quasi-Random Assignment of Trustees

- Chapter 7 trustees are assigned to cases via a blind rotation system
 - Morrison, Pang, and Zytnick (2019): evidence attorneys manip. Ch 7 trustee assignment
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 - ► Trustee fixed effect mitigates this concern, accounting for typical trustee behavior
- Chapter 13 Standing Trustees hired by local US Trustees Office
 - Each court has at most *several* Ch. 13 trustees at a given time; seem rotated
 - ► Time variation in local trustee race distribution ⇒ quasi-random assignment to filers
 - E.g., assume Florida is not more likely to have a Black Chapter 13 trustee at times when unobserved factors make dismissal less likely for Black filers

Plausibility of Random Assignment

1 Pairing of filer-trustee by race consistent with random assignment



Plausibility of Random Assignment

- 1 Pairing of filer-trustee by race consistent with random assignment
- **2 Balance Test:** filer characteristics do not predict trustee race



Outcome: Pr(Black Trustee); **Fixed Effects:** disposition year, district, filer ZIP, judge and trustee; **Obs:** 13.6M; **Clustering:** ZIP & Trustee (two-way, 95% confidence interval shown)

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► Table

Homophily Estimation Results

Coefficient



Fixed Effects: disposition year, district, filer ZIP, judge and trustee; **Obs:** 9.8M(Ch. 7), 3.6M (Ch. 13); **Clustering:** ZIP & Trustee (two-way, 95% confidence interval shown)

Homophily Estimation Results



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Quantifying Disparities Attributable to Bias

- Under...
 - Assumption 1: parallel disparities
 - Assumption 2: parallel accurate statistical discrimination
 - Assumption 4: non-white DMs weakly biased on average against Black filers
 - ... we can bound the share of disparities due to $\beta\mu$ -racial bias
- **Chapter 13:** $\tau_{13} = 0.10$ and 1 p = 0.83 imply $\delta_{13}^{\beta \mu} \in [0.08, 0.98]$ $\Rightarrow > 40\%$ of the 21 percentage point Chapter 13 dismissal disparity is due to bias

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- **Chapter 7:** $\tau_7 = 0$ and 1 p = 0.83 imply $\delta_{13}^{\beta \mu} \in [0, 1]$ \Rightarrow find no evidence of bias in Chapter 7

Black-White Dismissal Gap Correlates with Measure of Implicit Bias



Dismissal Rate vs Average IAT Score (County-Year Level)

Conclusion

Conclusion

- Black bankruptcy filers experience significantly higher bankruptcy dismissal rates
 - Observables explain most Ch 7 disparities, but only ~50% for Ch 13

• Formalize how **homophily** can detect and quantify $(\beta \mu)$ racial bias

• Black Ch 13 filers assigned to white trustees see **10% higher dismissal rates**

• Bias among bankruptcy DMs may limit Black households' access to debt relief

Thanks!

Race	Precision	Recall	F1-Score
Asian	0.67	0.60	0.63
Black	0.79	0.69	0.74
Hispanic	0.82	0.89	0.85
Other	0.40	0.04	0.07
White	0.87	0.95	0.91

Accuracy: % correctly predicted **Precision:** % of predicted identifications that are correct **Recall:** % of actual positives that are correctly identified **F1-Score:** harmonic mean of precision and recall

Prediction Success: Black



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Prediction Success: Asian



Pr(Asian) by Self-Reported Race

Prediction Success: Hispanic



Pr(Hispanic) by Self-Reported Race

Prediction Success: White



Pr(White) by Self-Reported Race

Prediction Success: Other





	(1)	(2)	(3)	(4)	(5)	(6)	(7)	
	Chapter 7 ($\mu = 0.023$)							
Pr(Black Filer)	0.030***	0.028***	0.029***	0.024***	0.024***	0.023***	0.006***	
	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(0.001)	(5e-04)	
Num.Obs.	39,002,506	38,985,463	38,985,463	38,985,463	38,985,463	38,985,463	11,977,436	
R2	0.002	0.004	0.008	0.280	0.283	0.289	0.055	
	Chapter 13 ($\mu = 0.559$)							
Pr(Black Filer)	0.209***	0.204***	0.173***	0.174***	0.169***	0.168***	0.106***	
	(0.003)	(0.004)	(0.002)	(0.002)	(0.002)	(0.002)	(0.002)	
Num.Obs.	14,122,752	14,114,534	14,114,534	14,114,534	14,114,534	14,114,534	4,487,022	
R2	0.019	0.064	0.097	0.406	0.417	0.424	0.305	
Disp. Year FE		\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	
District FE			\checkmark	\checkmark	\checkmark	\checkmark	\checkmark	
Filer ZIP FE				\checkmark	\checkmark	\checkmark	\checkmark	
Judge FE					\checkmark	\checkmark	\checkmark	
Trustee FE						\checkmark	\checkmark	
FJC Controls							\checkmark	

Clustering: ZIP; **Statistical significance:** 10%*, 5%**, 1%***

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Racial Disparities in Dismissal Rates (Controls)

Dismissal Rate Disparities



	Full Sample (1)	Chapter 7 (2)	Chapter 13 (3)
Pr(Black Filer)	0.044	0.009***	0.015
	(0.043)	(0.002)	(0.031)
1[Chapter 7]	-0.562***		
- · -	(0.070)		
Pr(Black Filer) x Pr(White Trustee)	0.128***	-0.003	0.101***
	(0.049)	(0.003)	(0.035)
Pr(Black Filer) x 1[Chapter 7]	-0.043		
	(0.044)		
Pr(White Trustee) x 1[Chapter 7]	0.111		
	(0.079)		
Pr(Black Filer) x Pr(White Trustee)	-0.130***		
x 1[Chapter 7]	(0.049)		
Observations	13,373,013	9,815,556	3,557,457
R2	0.460	0.052	0.306

Fixed Effects: disposition year, district, filer, ZIP, judge, and trustee; **Case controls:** 1[pro se], 1[prior filing], 1[nonexempt assets], 1[homeowners], 1[joint filing], ln(assets), debt/assets, % secured debt, ln(income), and income - expense gap; **Clustering:** ZIP and Trustee (two-way); **Statistical significance:** 10%*, 5%**, 1%***

Homophily: Additional Interactions

