

Can the Government Deter Discrimination? Evidence from a Randomized Intervention in New York City

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Racial discrimination persists despite established antidiscrimination laws. A common government strategy to deter discrimination is to publicize the law and communicate potential penalties for violations. We study this strategy by coupling an audit experiment with a randomized intervention involving nearly 700 landlords in New York City and report the first causal estimates of the effect on rental discrimination against blacks and Hispanics of a targeted government messaging campaign. We uncover discrimination levels higher than prior estimates indicate, especially against Hispanics, who are approximately 6 percentage points less likely to receive callbacks and offers than whites. We find suggestive evidence that government messaging can reduce discrimination against Hispanics but not against blacks. The findings confirm discrimination's persistence and suggest that government messaging can address it in some settings, but more work is needed to understand the conditions under which such appeals are most effective.

The passage of civil rights laws in the 1960s marked a turning point in the development of racial politics in the United States. Against the historical backdrop of slavery, Jim Crow, long-standing racial inequalities, and the marginalization of nonwhites, the legal prohibition of discrimination on the basis of race, color, or national origin signaled an important political shift toward a more egalitarian racial order (King and Smith 2005; Massey and Denton 1993). In addition to examining the political conditions leading to the passage of civil rights laws, scholars examining the politics surrounding the formal end of Jim Crow have focused on its aftermath. Scholars have documented how racial inequalities have persisted, and they have offered numerous

policy-centered, political economy explanations for their persistence (e.g., Alexander 2012; Forman 2012). Central to these explanations is the argument that racial disparities persist because racial discrimination is both pervasive and persistent despite established civil rights laws (Bobo, Kluegel, and Smith 1997; Dawson 2016; Dawson and Francis 2015). This calls into question the role of government in enforcing these laws, as well as the downstream consequences of government action (or inaction) for racial inequalities, the political economy of race, and the development of racial orders.

Understanding this puzzle requires pursuing two related lines of inquiry. The first asks: *To what extent does racial discrimination persist and why?*¹ The first-order challenge of

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1. We focus on understanding the incidence of discrimination and set aside questions about its causes, which have been widely studied in economics and sociology (e.g., Bertrand, Chugh, and Mullainathan 2005; Heckman 1998; Heckman and Siegelman 1993; Pager and Karafin 2009).

measuring discrimination levels is not trivial. For its potential targets, discrimination is difficult to detect because in many cases it is impossible to observe how a counterfactual individual would be treated in a market interaction. Going beyond firsthand accounts, relying on third-party reports is imperfect because they require unobtrusive observation and, as prior psychological research has documented, many individuals are unwilling to “see” or make attributions to discrimination even when confronted with direct evidence of it (Major and Dover 2016). To address this challenge, governments use enforcement audits to collect evidence of discrimination on a case-by-case basis, and both scholars and governments have relied on audit studies to measure the aggregate-level incidence of discrimination in employment and housing (see Bertrand and Duflo 2017; Edelman, Luca, and Svirsky 2017; Pager and Shepherd 2008; Turner et al. 2013). High-quality audits are, however, both costly and relatively rare. Most audit studies focus on discrimination in early stage market interactions but do not measure end-line outcomes (such as being offered a job or a housing unit), which are important to measure because they are key indicators of disparate impact.

The second, related line of inquiry asks: *Which government strategies are effective at reducing discrimination and why?* Despite an abundance of theoretical work on policy enforcement in general and research on the primacy of legal strategies in efforts to enforce antidiscrimination laws since their initial passage (see Cover 1995; Epp 1998; Frymer 2003; Rosenberg 1991; Skrentny 2002), to our knowledge no work exists that assesses whether governmental strategies to reduce discrimination are effective. Answering this second question is doubly difficult, as it requires measuring a hidden behavior and causal inferences around those behaviors.

In this article, we address both of these major questions and provide a first attempt to experimentally test the effects on racial discrimination levels of any government strategy to reduce them in the United States. We study the effectiveness of preemptive strategies to deter racial discrimination using official communication campaigns encouraging compliance with antidiscrimination law. We examine two commonly employed but understudied appeals: making the law itself salient and making the costs of violating the law salient. We work at a large scale that offers a good handle on discrimination levels, although, given fundamental measurement and inferential challenges, our treatment effect estimates are still measured with considerable uncertainty.

Our analysis examines discrimination and interventions in field conditions, in the context of government efforts to enforce fair housing law in the New York City rental market. We focus on the rental market as it is the segment of the

housing market in which reported discrimination is the most pervasive.² We partnered with the New York City municipal government to implement a large-scale field experiment that began in 2012 and lasted 20 months. We assess the effects on racial discrimination levels against black and Hispanic rental applicants (as compared to whites) of government appeals that are delivered via targeted and personalized phone calls to landlords and brokers who interact with confederates posing as rental housing applicants.³ The city randomly assigned to nearly 700 landlords either (1) a targeted live phone call from the city that drew attention to fair housing law and implicitly signaled increased government monitoring of housing agents (a “monitoring” condition), (2) a targeted live phone call with the contents of the monitoring message and additional information about the potential pecuniary punishments for discriminatory practices (a “punitive” condition), or (3) no call from the city (a pure control condition). The city then sent matched triples of confederates who varied by race (black, Hispanic, and white) to visit the same advertised unit—meeting in person with the same landlord—and to collect detailed qualitative notes on their interactions over the course of the housing search process. The enhanced audit design embedded in the experiment provides leverage above and beyond existing audit methodologies to measure both subtle and direct forms of racial bias that pervade the housing market. Additionally, in contrast to prior housing audit studies focusing on differential treatment in landlords’ responses to initial inquiries about rental listings, this study focuses on two outcomes occurring toward the end of the housing search process that are consequential for downstream housing inequalities: discrimination in receiving a callback following an in-person meeting with the landlord to view the listed unit and discrimination in receiving a rental offer from the landlord after the appointment.

The theoretical framework motivating the experiment is relatively simple. Drawing on existing research, we expect that the monitoring condition should decrease racial discrimination levels relative to a control, because receiving a personally targeted monitoring signal from a government known to actively enforce fair housing law is expected to reduce landlords’ perceived benefits from discriminating and to increase their perceptions of the probability of punishment if they discriminate. By contrast, existing research

2. Of the 18,978 housing discrimination complaints reported by private fair housing groups in 2014, 16,694 (or about 88%) occurred in the rental market (National Fair Housing Alliance 2014).

3. For simplicity, we refer to subjects, defined as landlords or brokers associated with a sampled rental housing ad and who interact with a matched team of testers, simply as “landlords.”

offers mixed expectations about the conditions under which coercive sanctions induce pro-social behavior, with some proposing possibly adverse effects; thus, we are agnostic about the expected direction of the punitive messaging effect.

We find evidence of substantial baseline levels of discrimination in the New York City rental housing market, particularly against Hispanics: they are 28% less likely to receive a callback (6.1 percentage points, 95% confidence interval, CI = [1.00, 11.19]) and 49% less likely to receive an offer for an apartment than whites (5.7 percentage points, 95% CI = [1.32, 10.16]). This result is striking when compared to discrimination levels reported by the 2012 US Department of Housing and Urban Development (HUD) audit of racial rental discrimination in the New York metropolitan area, which found a lack of discrimination against minorities in receiving callbacks (discrimination in receiving offers was not studied).⁴ We show that racial discrimination in fact persists in the New York City rental market. We also find suggestive evidence that treatment messages reduce levels of discrimination in receiving callbacks, although these effects are statistically weak and are observed only for discrimination against Hispanics. There is no evidence that discrimination against blacks is affected by these government messaging interventions. We also find no evidence that the punitive messages are more or less effective than simple monitoring messages. Our results suggest that a bundled governmental messaging strategy that invokes fair housing law and makes salient the costs of punishment for lawbreakers can be effective at reducing racial housing discrimination, but not under all circumstances. As a whole, this study makes the case that evaluating government efforts to reduce racial discrimination is central to understanding the political economy of race and motivates the need for additional future research on the conditions under which government efforts to enforce antidiscrimination laws and reduce discrimination are effective.

THEORETICAL FRAMEWORK AND EXPECTATIONS

Government strategies to reduce racial discrimination in the United States

Government strategies to enforce antidiscrimination law may be classified into one of two broad types: reactive or preemptive. Reactive strategies such as “fire alarm” and “police patrol” models of policy enforcement (McCubbins and Schwartz 1984) are those in which governments investigate and litigate

potential discriminators as an ex post response conditional on citizens and governments, respectively, observing and reporting discrimination in the first place. Given the challenges with observing discrimination (as previously discussed) and the expectation that pursuing litigation may incur disproportionate financial and psychological costs on the targets of discrimination relative to defendants (Donohue and Siegelman 1991; Feagin 1991; Galanter 1974), reactive strategies are limited in their ability to identify, much less punish, most who discriminate.

In contrast, preemptive strategies are those where official messaging campaigns attempt to deter discrimination ex ante, by advertising existing antidiscrimination laws and making various appeals that attempt to induce compliance with the law. Such campaigns complement reactive strategies and are attractive in part because, if effective, they could offer a lower-cost method to induce behavioral change (as compared to deploying an intensive enforcement audit program). Existing government communication campaigns typically involve bundled messages and appeals that aim to reach multiple potential audiences including the targets of discrimination, discriminators, and third-party observers. Evaluating existing bundled government campaigns does not provide leverage to understand why enforcement efforts are effective, however, because it would not be possible to infer whether potential lawbreakers decide to comply with the law because they perceive the costs of violating the law to outweigh the benefits (Wilson 1980). To disentangle the effects of different governmental appeals on individuals whose behavior fair housing laws ultimately aim to change, we focus on testing whether governmental appeals that directly target potential discriminators (i.e., landlords and brokers) and that are designed to alter their expected returns to discriminating are effective at inducing behavioral change.

When do messaging campaigns reduce discriminatory behavior?

Under what conditions might government messaging campaigns be effective at reducing discriminatory behavior? Existing research undertheorizes why official appeals would be effective at reducing discrimination. We develop a simple decision-theoretic model to clarify the basic logic. In the context of a housing market interaction, the individual of interest is the landlord or broker who interacts with housing applicants and may potentially engage in discriminatory behavior. Discrimination (as an individual-level behavior) simply refers to the differential treatment of two individuals who vary on a relevant attribute such as race. This definition is agnostic as to why people discriminate, which may arise because

4. The estimates from both samples condition on matched testers making an appointment, but the samples also differ. The 2012 HUD study had a larger initial sampling frame (including northern New Jersey and units found through non-Craigslist means), whereas our sample is larger yet only includes New York City rental listings from Craigslist.

people hold group-specific prejudices and a taste for discrimination (Becker 1957) or because people engage in statistical discrimination by relying on average group stereotypes to make inferences about a person perceived to belong to that group (e.g., Aigner and Cain 1977; Arrow 1973; Phelps 1972). Let individual i 's utility from discrimination given government policy x be expressed as $u_i(x) = b_i(x) - p_i(x)c_i(x) + e_i$, where b_i denotes marginal benefits from discriminating, p_i the subject's expectation of punishment when she discriminates, c_i the expected severity of the punishment, and e_i individual heterogeneity. If $b_i(x) > p_i(x)c_i(x)$, then i will discriminate.

Building on this framework, we aim to estimate both the baseline level of housing discrimination against black and Hispanic renters as well as the causal effects on racial housing discrimination of governmental messaging that encourages landlords and brokers to comply with fair housing law. We investigate these quantities separately for blacks and for Hispanics because in multiracial contexts, there are multiple dimensions of social difference that are potentially salient, and the nature of bias against a racial group is expected to vary as a function of how each group is perceived to deviate from a majority reference group (e.g., Dovidio et al. 2010; Pehrson, Vignoles, and Brown 2009).⁵ We examine two distinct types of policy: a monitoring signal that invokes the law and implicitly signals increased government monitoring of housing agents, and a punitive appeal that additionally primes the costs of discrimination. For practical reasons—in particular to satisfy our implementing partner's preference for treatment realism and to avoid depleting statistical power—the experiment is designed to test realistic, bundled treatments but does so at the expense of testing specific parameters. We are cognizant that this is a major limitation of this study and recommend future work to test treatments that target each parameter separately. This framework is nevertheless useful to develop expectations about why the appeals we test would affect discrimination.

We test the hypotheses that sending monitoring messages alone decreases rental housing discrimination rates against blacks and Hispanics (hypothesis family **H1**). There are two possible reasons for why it might. First, simply contacting landlords and making the law salient may raise expectations of sanctioning, p , even if this is not explicitly invoked and no information on sanctions is provided. Given the context of

this study, the mode of treatment delivery, and the particular messenger of the treatment appeal, we argue that this is plausible in this experimental setting. Because the treatment calls in this study are personalized and targeted and because they are sent by a city government known to have the capacity and willingness to monitor and enforce fair housing law, receiving such a call provides a strong and credible signal to the landlord that the city is already monitoring them. Thus, receiving any targeted call from the city may increase p above zero.

Increasing the salience of injunctive norms about fair housing in the landlord's thinking may also reduce b . Simply invoking the law may activate compliance norms and may also render compliance focal (McAdams and Nadler 2005; Tyler 2006). For discrimination in particular, Mendelberg (2001) and others have highlighted how explicit priming of discrimination considerations can result in less discrimination (relative to implicit priming) due to invoking social non-discrimination norms. If housing agents perceive discrimination to be a descriptive norm, then these perceptions may drive beliefs that the law is illegitimate and result in less compliance (Tyler 2004, 2006). But appeals to injunctive norms have been theorized to be effective at crowding out descriptive norm perceptions used to justify noncompliance (Weaver 2015). Our estimates of the monitoring effect thus capture joint effects on p and, to the extent that preferences are not fixed, b .

In addition, we study the effect of altering c by assessing the effect of sending "punitive" messages, relative both to the control condition (thus the combined effect of monitoring and punitive content) and to the monitoring condition (thus the additional effect of punitive content as compared to the monitoring condition). Thus, in practice we assess the hypotheses that sending punitive messages decreases rental housing discrimination rates against blacks and Hispanics relative to the baseline condition and to a monitoring condition (hypothesis family **H2**).⁶ Although we have clear expectations about the marginal effect of costs, c , we are cognizant that our punitive treatment may also have the effect of altering b , relative to the monitoring treatment.

Existing research suggests mixed expectations about whether amplifying the salience of the severity of punishment will be effective at reducing discrimination. On the one hand, theoretical work on the politics of inducing behavioral policy compliance argues that when individuals face insufficient incentives to comply or perceive complying as costly, a potentially effective strategy to induce compliance involves increas-

5. Our study therefore addresses the need for additional research on discrimination against Hispanics (Dovidio et al. 2010) and the need to study discrimination against multiple groups in the same study to understand discrimination in a multiracial context, which is rare in the literature.

6. The latter comparison can be interpreted as the pure effect of explicitly priming punishment.

ing and making salient the costs of noncompliance such that they are perceived to be greater than the costs of compliance (e.g., Dixit 2006; Hadfield and Weingast 2014; Weaver 2014, 2015). Prior research evaluating the effectiveness of similar official communication campaigns has shown that appeals highlighting the costs of noncompliance can, under certain circumstances, induce greater compliance with the law in domains such as paying taxes (Blumenthal et al. 2001; Iyer, Reckers, and Sanders 2010; Slemrod, Blumenthal, and Christian 2001; but see Dunning et al. 2015) and paying delinquent fines (Haynes et al. 2013). On the other hand, a growing line of social psychological research challenges this expectation and argues that coercive sanctions to induce behavioral change may backfire and instead crowd out social norms needed to motivate pro-social behavior (e.g., Bowles 2016; Cardenas, Stranlund, and Willis 2000; Gneezy and Rustichini 2000). Thus, while we gather direct evidence on the effect of the punitive treatment, this can be interpreted as the effect of costs, c , only under the assumption of no differential effects, relative to the monitoring treatment, on b and p .

EXPERIMENTAL DESIGN

Partnering with the City of New York, we designed and analyzed data from a field experiment in which the city randomly sent targeted messages to landlords associated with a specific advertised rental unit, which is pursued by a trio of matched testers who vary by race. The matched audit provides leverage to measure racial discrimination, which we operationalize as differential landlord behavior toward testers by the tester's race for the same advertised unit. This section describes the experiment's setting, design, implementation, and analysis.

Study context

New York City is a useful political setting for testing the effectiveness of governmental campaigns encouraging compliance with fair housing law. It is well known for having one of the strongest antidiscrimination laws in the country (Gurian 2005). The agency enforcing the city's antidiscrimination law, the New York City Commission on Human Rights, is also well known among real estate professionals for having the capacity and willingness to enforce fair housing law. Thus, receiving any targeted and personalized treatment phone message from this city government is arguably perceived as a credible signal of increased government effort to monitor landlords and enforce fair housing law. We focus on rental listings posted on the online classified advertising site Craigslist. While Craigslist is one of multiple sources of classified rental advertisements, it is one of the primary forums used to post and pursue rental listings in general and in New York City.

Audit design

We briefly summarize how rental listings are sampled from Craigslist and how they are pursued by matched auditors.⁷ On each day of the study's implementation, a set of rental housing ads from the current day was selected from Craigslist using an automated script. First, using keywords, a list of "likely discrimination" (LD) ads were identified and selected with 100% probability, with the goal of increasing statistical power if baseline discrimination levels were low.⁸ Second, among the remainder of ads posted on Craigslist that day (excluding those identified in the LD search), a sample of ads was randomly drawn in a way that was representative of the distribution of advertised vacant rental housing stock by New York City borough.⁹ Only advertisements that invited housing seekers to reply by phone were pursued.¹⁰

Testers responded to sampled ads posing as individuals interested in renting the listed apartment. A project manager prescreened all sampled ads against a master database of prior audits to ensure that there were no duplicate landlords or brokers in the sample of ads ever pursued by testers. We refer to all ads pursued by testers as the "audit sample." Each ad was pursued in a randomized order by a matched team of three testers of the same gender who varied by race: one white, one black, and one Hispanic.¹¹ By extensively controlling for aspects of testers' assumed biographies within each trio, we employ a design-based approach to reduce the likelihood that observed racial discrimination levels are driven by the traits of specific auditors. Assigned biographies logically consistent with the rental price and size of the advertised unit were automatically generated at the time of ad sampling.

Upon reaching an individual when replying to an ad, testers were instructed to provide limited information about themselves over the phone, including their assumed name, interest in pursuing and renting the unit, their availability for

7. Full details about the sampling, audit, and field procedures are provided in the appendix.

8. All LD ads were pursued ($n = 156$), and of these, 44 were associated with landlords admitted into the experiment. To address concerns that their inclusion increases discrimination levels in the control group and decreases effect magnitudes if landlords in the LD subsample are more likely than non-LD landlords to discriminate, we find that the main results are not materially affected when the LD subsample is excluded. See the appendix for full results.

9. These are based on the distribution of ads by borough as identified in a pilot study: 35% Manhattan, 30% Brooklyn, 20% Queens, 10% Bronx, and 5% Staten Island.

10. In a pilot study, we found that initial contact rates for e-mailed inquiries were extremely low.

11. We matched testers' assumed credit score range, income level, household composition, occupation, perceived employment stability, gender, interpersonal skills, and perceived age.

an in-person viewing, and their financial qualifications to rent.¹² Testers were instructed to schedule an in-person appointment on the same day at the earliest convenience of the individual with whom they were speaking. If testers were asked about other biographical attributes, they volunteered that information accordingly.

When all three testers in a matched team successfully secured an appointment to view the same advertised housing unit with the same landlord, the landlord associated with that unit was admitted into the experiment (i.e., the “experimental sample”) and randomly assigned to a treatment condition. We refer to the landlord-housing unit (and thus the unit of randomization) as a “case.” Testers then made their individually scheduled appointments, viewed the unit, interacted with the subject, and recorded their interactions afterward. Testers were blind to treatment, but they were not blind to the city’s interest in assessing discrimination as one of numerous characteristics of the rental market. As such, they were extensively instructed and coached not to fish for particular reactions, not to let their personal opinions about landlord behavior interfere with their ability to continue interactions during the audit, and not to raise suspicions of an audit.

Randomization procedure and definition of treatments

Landlords admitted into the experiment were randomly assigned to one of three conditions: a control condition where no message was sent, a monitoring messaging condition, or a punitive messaging condition. A block randomization procedure was used in which blocks were defined by the ad’s stratum and by treatment regime.¹³ Table A1 (tables A1–A23 are available online in the appendix) shows the distribution of cases across blocks and arms.¹⁴

Treatment scripts contain official language that the city uses to communicate aspects of fair housing law to the pub-

12. Testers had an assumed first name and an assumed last name for the duration of the study. Both were randomly drawn from a database of names tagged with racial and ethnic identifiers (see the appendix for details). While testers were screened at the time of hire such that their manner of speaking did not strongly signal a particular racial identity over the phone, testers’ assumed names did contain a signal of their race (Butler and Homola 2017; Fryer and Levitt 2004).

13. Strata were a New York City borough or the LD oversample. The treatment regime is defined as a distinct design and randomization procedure. See fig. A2 for details.

14. We verified the randomization was valid using randomization inference. The probability of obtaining a log-likelihood statistic (from a multinomial logistic regression of treatment assignment on 122 pretreatment covariates) at least as large as the observed test statistic is $p = .97$. Balance tables are presented in the appendix.

lic.¹⁵ In both the monitoring and punitive conditions, a city employee delivered the assigned appeal via a personalized, targeted phone message to the landlord. Calls were sent about two hours after testers successfully scheduled appointments over the phone and about two hours before the first scheduled appointment. In the monitoring condition, the treatment script verified that the subject was on the line, informed the subject that the call was from the New York City Commission on Human Rights “as part of an ongoing informational campaign to remind landlords and brokers of their obligations under fair housing law,” and provided the commission’s web address for more information. While the monitoring message does not provide specific information about what exactly is illegal and is in fact designed to avoid priming subjects to think about racial discrimination in particular, our prior is that most of the subjects already knew about fair housing law in the absence of treatment.¹⁶ Thus, the monitoring message can be interpreted as an intensive intervention that makes salient injunctive norms surrounding fair housing law simply by invoking it. In the punitive condition, the treatment script was the same as the monitoring treatment script but includes, before providing the commission’s web address, the following advisement emphasizing the punitive power of the state and the potential pecuniary costs of violating the law: “It is illegal to discriminate against a person seeking housing due to their membership in a protected class. If you are found to have broken the law, you may be ordered to pay damages, provide reasonable accommodation, or incur civil penalties of up to \$250,000.”

Data and measurement

We combine the following data: (1) scraped data on advertised rental listings, (2) automated assumed biographical and treatment assignment data, (3) data on treatment delivery and compliance, and (4) detailed field reports testers recorded about their interactions with landlords before the visit (pretreatment), during a housing unit visit (posttreatment), and after the visit (posttreatment).¹⁷ We construct case-level measures of net discrimination, which is defined as the difference in favorable treatment between the majority and minority group tester in any matched pair. We apply this measurement strategy to construct two objective measures of discrimination, which we prespecified as our primary posttreatment outcome measures:

15. See the appendix for the full text of the treatment scripts.

16. This is because brokers, who comprise nearly 85% of experimental subjects, are required to demonstrate knowledge of fair housing law to be licensed in New York and because Craigslist shows advisory language stating that discrimination is illegal to those posting rental ads.

17. Full details about data collection and measurement procedures are provided in the appendix.

differential treatment by race in landlord or broker efforts (i) to call back and follow up with testers after the appointment and (ii) to offer the unit to the tester. These measures are computed for each majority-minority pairing (white-black, white-Hispanic, and black-Hispanic) and can take three values at the case level: -1 if only the minority tester is treated favorably, 0 if both the minority and majority testers are treated equally, and 1 if only the majority tester is treated favorably.¹⁸ When examining average levels of the net measure, 1 means 100% net discrimination against the minority group, -1 means 100% net discrimination against the majority group, and 0 means that the two groups are treated equally.¹⁹ We also construct pretreatment measures of discrimination that occurs over the phone between the time testers initially contact the landlord when inquiring about the ad and the time of random assignment. We measure early stage discrimination for all cases testers pursue to describe baseline discrimination levels and to include as covariates in the experimental analysis.²⁰

Sample definitions

The study was in the field from April 13, 2012, to December 20, 2013. We focus on two key samples for the analysis: the audit sample, which contains 2,711 cases, and the experimental sample, which contains 653 cases. Figure A1 (figs. A1–A7 are available online in the appendix) presents a flow diagram summarizing the sample filtering procedure, and figure A2 summarizes the cumulative number of cases admitted into the experimental sample over this period and by treatment regime. The characteristics and geographic distribution of the rental housing stock in our samples appear to be broadly similar to the stock in the New York City rental housing market during this period.²¹

18. If the landlord did not honor the appointment, the net discrimination measure is coded 0 (i.e., both in any pair were treated the same as neither received a callback nor an offer).

19. A concern is that estimates of discrimination levels are driven by the composition of testers in each racial group. To address this, we estimate a nonnested hierarchical model regressing a landlord-tester level indicator for favorable treatment on pretreatment covariates and tester, tester race, and tester team gender random effects. We estimate this model for each of the objective outcome indicators among the control group and among the experimental sample. The estimated variance of the varying tester intercepts is negligible; thus, we infer that the composition of testers in each racial group does not drive discrimination estimates. See the appendix for details.

20. Early stage discrimination indicators include differences in making initial contact; scheduling an appointment; how many biographical attributes the landlord asks about; the number and share of attributes discussed where testers receive positive, negative, neutral, or skeptical feedback; and whether any negative or skeptical feedback is received.

21. See the appendix for summary statistics of the housing stock characteristics of listed units by sample.

Estimation and inference

We use both design-based and model-based approaches to estimation and inference, with both approaches producing near-identical results.²² For design-based (nonparametric) inference, we estimate sample intent-to-treat (ITT) effects of government messaging on discrimination levels as the weighted average across blocks of differences in net discrimination levels between treatment groups within blocks, where weights are given by block size.²³ For this nonparametric analysis, standard errors are calculated using the conservative weighted Neyman estimator. In addition, we estimate effects using the following linear model:

$$Y_{ib} = \alpha_0 + \beta_1 T_{ib} + \gamma_b + u_{ib}, \quad (1)$$

where i indexes landlords and b indexes the experimental block; Y is the net discrimination outcome measure; T is treatment assignment, variously defined when comparing mean outcomes between monitoring versus control, punitive versus control, or punitive versus monitoring; γ is a full set of block fixed effects; and u is a disturbance term. We estimate equation (1) on the subset of the data assigned to each pair of treatment arms being compared; β_1 is the effect of the treatment group relative to the comparison group. For this analysis we again use inverse propensity weights. We calculate p -values corresponding to a one-sided test of the null hypothesis of equality of means for the monitoring-control and punitive-control comparisons. We also calculate p -values corresponding to a two-sided test of the null hypothesis of equality of means for the punitive-monitoring comparison and for all analyses involving net discrimination against Hispanic (vs. black) testers because we have no strong priors about the expected direction of racial bias and treatment effects on racial bias when comparing the experiences of black and Hispanic testers.

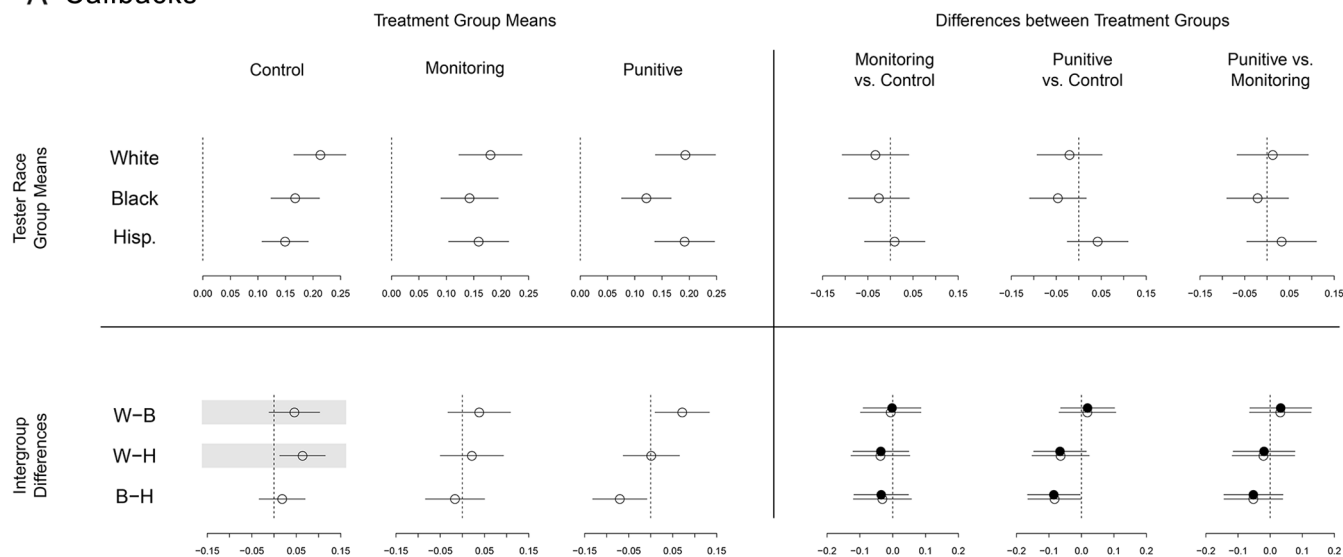
RESULTS

Figure 1 summarizes the main findings and shows, separately for each outcome measure, (1) levels of favorable treatment for different racial groups (*top-left quadrant*), (2) differences in favorable treatment rates between groups (i.e., net discrimination levels) by treatment assignment (*lower left*), (3) differences in favorable treatment rates across treatment

22. Additional analyses are described and reported in the discussion and in the appendix.

23. We observe some noncompliance because some landlords assigned to a call hung up midcall or because the city staffer delivering calls could not reach the landlord before tester appointments. Because the estimated share of compliers is relatively high (between 71% and 81%), complier average causal effect estimates are similar to ITT estimates and are shown in the appendix.

A Callbacks



B Offers

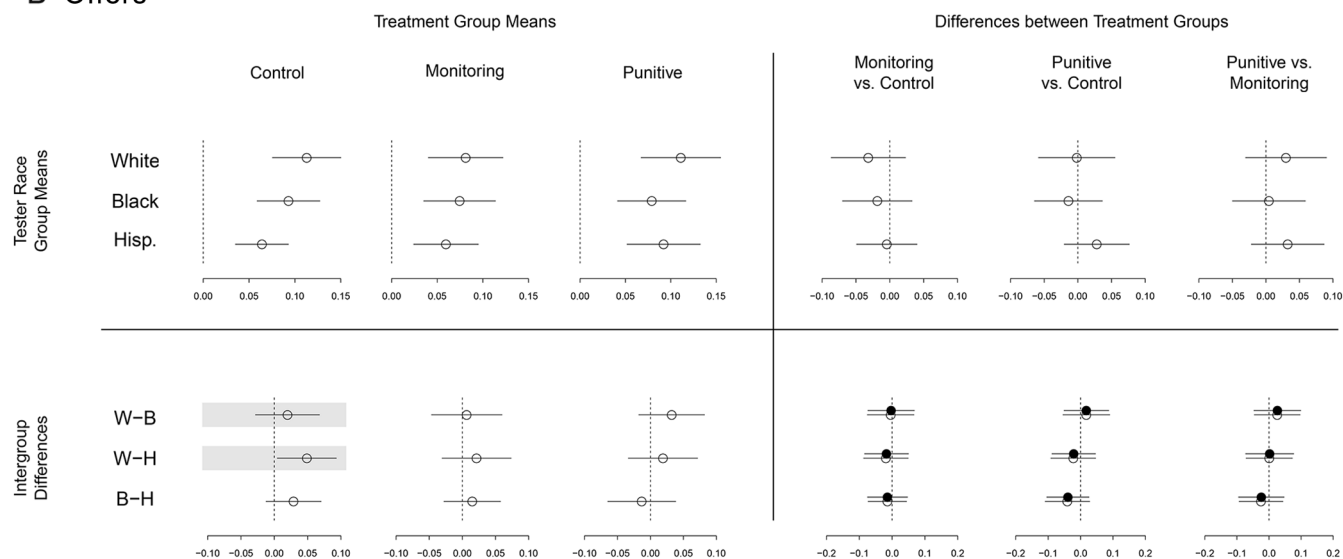


Figure 1. Main results on discrimination levels and treatment effects. *A*, Net discrimination in receiving callbacks; *B*, net discrimination in receiving offers. Each panel shows levels of favorable treatment for different racial groups (*top-left quadrant*), differences in favorable treatment rates between groups (i.e., net discrimination levels) by treatment assignment (*lower-left quadrant*), differences in favorable treatment rates across treatment conditions for the same group (*top-right quadrant*), and the effects of treatment assignment on net discrimination levels relative to the control or monitoring comparison group (*lower-right quadrant*), with weighted nonparametric estimates shown using open markers and inverse probability weighted regression estimates adjusted using block fixed effects shown using filled markers. Lines indicate 95% confidence intervals. Our main quantities of interest are highlighted in gray.

conditions for the same group (*top right*), and (4) the effects of treatment assignment on net discrimination levels relative to the control or monitoring comparison group (*lower right*).²⁴ We focus our discussion on the lower-left and lower-right quadrants in particular, which present estimates of baseline

net discrimination levels and of ITT effects of messaging on net discrimination levels, respectively. We also draw attention to our most important findings in the figure by highlighting key estimates.

Baseline discrimination levels

First we assess baseline levels of discrimination in the outcome variables by examining the control group mean levels

24. Estimates corresponding to the information presented in fig. 1 are shown in the appendix.

of net discrimination, which are defined as the control group mean difference in favorable treatment rates between groups, as shown the lower-left quadrants in figure 1. Here we find statistically significant differences between Hispanic and white testers. We estimate that Hispanic testers were less likely than white testers to receive a callback from a landlord or broker—in 14.9% of cases compared to 21.3%, a difference of 6.4 percentage points ($p = .014$). They were also less likely to receive an offer for an apartment—in 6.4% of cases compared to 11.3%, a difference of 4.9 percentage points ($p = .031$). We find smaller, and statistically weaker, differences between black and white testers. We estimate that black testers were less likely than white testers to receive a callback from a landlord or broker—in 16.8% of cases compared to 21.5%, a difference of 4.6 percentage points ($p = .115$). Black testers were also less likely to receive an offer—in 9.3% of cases compared to 11.3%, a difference of nearly 2.0 percentage points ($p = .418$).

We can compare baseline levels of discrimination in callbacks to estimates of differential treatment in receiving a follow-up from the 2012 HUD audit of ethnic and racial discrimination in rental markets.²⁵ Our results run counter to the New York metropolitan area estimates from the 2012 HUD audit, which found that Hispanic testers received a follow-up from agents more frequently than white testers (in 6.9% and 5.4% of cases, respectively, a difference of -1.5 percentage points, $p = .804$) and that black testers received a follow-up from agents more frequently than white testers (in 10.2% and 3.9% of cases, respectively, or a difference of -6.3 points, $p = .096$; Turner et al. 2013, 157–58). These differences in apparent baseline levels of discrimination might be due to two factors. First, the two estimates may be comparing discrimination baseline levels for two different populations in the New York area rental market. The HUD estimates for the New York City area correspond to the New York/northeast New Jersey metropolitan area and not to New York City specifically. Our estimates are specific to New York City, and in particular to landlords and agents who post ads on Craigslist and who schedule appointments with all three testers pursuing the listing. Second, there is less variability in our estimates than in the New York area estimates from the HUD audit since the sample size of our control group is nearly twice the size of the entire New York HUD audit. When comparing our estimates to national estimates of discrimination in follow-ups from the 2012 HUD study, the estimates are similar in that black and Hispanic testers were less likely to receive follow-ups from agents than white testers: 10.5% for black testers versus 11.0% for white testers, a

difference of 0.6 percentage points that is not significant at the 0.05 level; and 7.9% for Hispanic testers versus 11.2% for white testers, a difference of 3.3 percentage points that is significant at the 0.1 level (Turner et al. 2013, 44, 47).

Does government messaging reduce racial discrimination?

Next, we report estimates of the effect of messaging on net discrimination levels, as shown in the lower-right quadrants in figure 1. This quadrant presents weighted nonparametric estimates (open markers) and regression estimates adjusted using block fixed effects and inverse probability weighting (filled markers; our preferred specification) with 95% CIs.²⁶

When compared to a pure control condition, sending a monitoring message decreases net discrimination against both black and Hispanic testers (vs. white testers) across the objective outcome measures, but mean effect estimates are substantively negligible and not statistically distinguishable from zero. Sending a monitoring signal decreases net discrimination against black testers in receiving a callback (-0.006 , $p = .897$) and in receiving an offer (-0.004 , $p = .903$). For Hispanic testers, sending the monitoring message decreases net discrimination in receiving a callback (-0.037 , $p = .412$) and receiving an offer (-0.019 , $p = .586$).

The results on punitive messaging point generally to reductions in discrimination against Hispanics but increases in discrimination against blacks. Sending a punitive message increases net discrimination against black testers (vs. white testers) across the outcome measures when compared to the control, but none of these effects are statistically significant at the 5% level. Sending the punitive message reduces the likelihood of receiving a postvisit follow-up callback from the landlord (0.02 , $p = .679$) and reduces the likelihood of receiving a postvisit offer (0.018 , $p = .624$). Punitive messaging instead decreases net discrimination against Hispanic testers in receiving a postvisit callback (-0.066 , $p = .056$) and in receiving a postvisit offer (-0.023 , $p = .522$). The effect of punitive messaging on net discrimination in receiving a postvisit callback is just shy of significance at 5% ($p = .056$). Notably, sending a punitive message worsens outcomes for black compared to Hispanic testers across the outcome measures. Punitive messaging decreases net discrimination against Hispanic testers relative to black testers in receiving callbacks (-0.083 , $p = .049$) and in receiving offers (-0.040 , $p = .247$). Note that of these two estimated treatment effects, only the effect on callbacks is statistically significant at the 5% level ($p = .049$).

25. No comparable measure of discrimination in offers exists in the HUD audit study.

26. We focus on the latter in our exposition of results, because both yield qualitatively similar results.

We next examine the comparative effectiveness of punitive versus monitoring messaging. It is clear from figure 1 that there is no statistically distinguishable difference in the effectiveness of the two treatments. We find mixed results when comparing outcomes between the two treatment groups—black testers do slightly worse when interacting with landlords in the punitive condition as compared to the monitoring condition, while Hispanic testers do slightly better. All *p*-values for these objective outcomes range from 0.261 to 0.984.

Hidden discrimination: Estimates using subjective measures

Our primary analysis focuses on the causal effects of government messaging on two objective measures of racial discrimination in the housing market. Focusing on objective measures is important in part because discrimination in the housing market can be relatively invisible (Fair Housing Justice Center 2013; Freiberg 2015). We also have access to a set of subjective measures of discrimination that capture, among other aspects of the housing search process, subjects' perceptions of steering. As such, they reflect both the behavior of the landlords, which may be a function of treatment conditions, and the dispositions of confederates to notice and report indicators of discrimination, which is not subject to treatment conditions but may be associated with racial group membership. The measures, while rich, are also flawed in that they suffer from substantial missingness (on the order of 25%–30%).²⁷ While missingness is not correlated with racial groups or treatment arms, it is still plausibly related to potential outcomes: testers may be less likely to report if they felt either little or substantial discrimination.

With this caveat in mind, in figure A3, we report messaging effects on an index measure of testers' subjective perceptions of favorable treatment and a net discrimination measure using the subjective index.²⁸ Although on the objective measures there is strong evidence for discrimination against Hispanics, this does not appear at all in the subjective measures. Moreover, while there is some evidence for discrimination against blacks from the objective measures, for the subjective measure, blacks report, if anything, better treatment than whites. Our results are consistent with the dominant viewpoint in the field, but they also have implications for the

generality of our findings. The difficulty of identifying discrimination based on subjective perceptions may be one factor that makes the punitive threat of law relatively weak, even if there are normative benefits. Different effects may arise in areas where identifying violations is easier.

DISCUSSION

Research on the political economy of race in the United States has shown that racial discrimination plays a central role in perpetuating racial inequalities, but it has long overlooked questions about whether government efforts to reduce discrimination have been effective. We therefore conduct the first experimental test of two common government strategies to preemptively induce behavioral compliance with the law in the domain of fair housing. Partnering with the City of New York, we assess whether government campaigns that invoke fair housing law and make salient the costs of violating the law reduce racial discrimination levels in the New York City rental market. The design offers a strong test because it (1) individually targets appeals at potential discriminators using live calls, (2) measures outcomes soon after the treatment is delivered, and (3) occurs in a policy context in which government's monitoring and enforcement signals are credible.

First, in the absence of government intervention, we find strong evidence of discrimination in the New York City rental market on outcomes that are consequential for racial housing inequality. In the control group, Hispanics are less likely than blacks, who are themselves less likely than whites, to receive favorable treatment from landlords. In terms of the objective outcome measures, we find strong evidence of discrimination in the New York City rental market, particularly against Hispanics. They are 6.4 percentage points less likely to receive a callback and 4.9 percentage points less likely to receive an offer for an apartment than whites. These findings stand in contrast to estimates of discrimination levels from a comparable audit study of the New York metropolitan area rental market in 2012, which reports no discrimination against blacks or Latinos in receiving a callback. We also find suggestive, but not strong, evidence that punitive government messaging campaigns that both invoke the law and make salient the costs of violating the law can reduce net discrimination in receiving a callback from a landlord following an appointment, but only for Hispanics. We find no statistically distinguishable difference between the effects of punitive and monitoring messaging relative to a control.²⁹ We assess the robustness of our main

27. We refer the reader to the appendix for an analysis of missingness on subjective measures.

28. We use an index measure to avoid a multiple comparisons problem. We conduct and report this analysis to maintain fidelity to our preanalysis plan and to situate our main results in a broader policy context. Details about the procedures related to this analysis are shown in the appendix.

29. Employing a conservative Bonferroni approach to multiple comparisons across treatments, outcomes, and groups, we fail to reject the null that these treatments make no difference.

results (from preregistered analyses) to alternative estimation strategies that, at least in principle, improve efficiency; our main results are not materially affected by the choice of estimator.³⁰

Despite the lack of statistical significance, the magnitudes of the estimated messaging effects we observe in this study appear to be substantively significant both in relative terms and in terms of the mean level of discrimination under treatment. In relative terms, the effects are very large, albeit from a small baseline.³¹ Punitive messaging reduces discrimination against Hispanics relative to whites by 109% and relative to blacks by 126%. Not only is there a large relative decrease in net discrimination against Hispanics, but the sign of the predicted group mean changes from positive to negative, suggesting more favorable treatment for Hispanics relative to whites and blacks under the punitive messaging condition. Alternatively we can adduce the substantive meaning of effects by interpreting the mean discrimination level under treatment. Our estimates of discrimination levels on key measures drop to zero under the punitive condition. This creates an inferential puzzle for policy makers as well as a need for further experimental replication, which we discuss below as well. In the remainder of the article, we address concerns regarding the internal and external validity of our findings, implications for policy makers, and directions for future research.

Addressing internal validity concerns about spillovers

The main threat to internal validity arises if the noninterference assumption is violated, which could occur if landlords assigned to a messaging condition communicate with other landlords (who are assigned to other arms) about the treatment message they received. We use two main strategies to show that such interference is unlikely given the sheer size of the New York City rental market and the sampling procedures.³² First, we estimate that the probability a landlord or broker enters the audit sample is between 3.2% and 15% and that the probability a landlord or broker enters the experiment sample is between 0.77% and 3.6%. Second, we char-

acterize the experimental sample as a very small random sample of landlords and brokers in the New York City rental market. Because the actual number of active landlords and brokers in the New York City rental market is unknown, we estimate a conservative upper bound to proxy the denominator. We infer that the experiment sample must be far less than 2% of the estimated population of landlords and brokers in the New York City rental market. Taken together, these estimates suggest a low probability of between-subject interaction and make a strong *prima facie* case for minimal interference.

Addressing external validity concerns

Next, we address concerns related to external validity deriving from the ways study samples are constructed. Although care was taken to sample ads randomly from the universe of Craigslist rental listings, there are several scope conditions that limit the generalizability of our findings. First, our results are limited to ads placed via public listings on Craigslist, which is a subset of advertised market-rate rental units and excludes discrimination involving nonpublic tenant search processes and discrimination in the subsidized rental market. This would underestimate discrimination levels under the assumption that landlords who post rental ads on Craigslist are likely to be less discriminatory than those who select into nonpublic modes of advertising.³³

Second, our findings are also limited to landlords and brokers who require housing seekers to reply to rental ads by phone.³⁴ This is a scope condition on our findings, and future research should explicitly assess whether findings differ by the mode of initial contact. Third, a possibly relevant factor is that Craigslist employs its own antidiscrimination messaging, which is shown to users before posting and acknowledges federal, state, and city prohibitions on discrimination.³⁵ If this messaging deters would-be discriminators from advertising on Craigslist, then this would bias our estimates of baseline discrimination levels downward. If it simply changes the ways in which landlords and brokers advertise, then this makes the evidence for discrimination all the more striking, as the discrimination we uncover cannot

30. First, using all the data we estimate models using binary treatment indicators, block fixed effects, and inverse probability weights, thus allowing for possible efficiency gains from tighter block effect estimation. Second, we use a lasso regression as a principled method for covariate selection and then estimate covariate-adjusted ITT effects. See the appendix for full results.

31. Interpreting treatment effects as relative differences offers an alternative substantive interpretation and is customary in experimental analyses. See the appendix for full results.

32. Methodological details and additional analyses are in the appendix.

33. This is consistent with research documenting the role of segregated informal social networks in employment and housing search processes (Loury 2001; Pager and Shepherd 2008).

34. We estimate that 55.8% of New York City rental listings on Craigslist require contact by phone. See the appendix for details.

35. The message states, in part: "It is illegal to discriminate in the sale, rental or leasing of housing because of a person's race, color, creed, national origin, sexual orientation, marital status, familial status, or religion."

be readily detected from simple text analysis.³⁶ It is also possible that Craigslist's messaging, by having effects of its own on discrimination and reducing the baseline level of discrimination, reduces the magnitude of the effect of government messaging. If this is the case, then our results should be interpreted as evidence of the effects of government messaging in a context in which nongovernmental messaging also exists.

Fourth, our experiment applied treatment only after appointments were scheduled, to avoid differential attrition. This might have the effect of limiting the generalizability of our analysis to encounters in which early stage discrimination was absent. We can address this concern by assessing discrimination toward testers occurring over the phone before randomization—that is, across the full audit sample. Table A2 summarizes these findings and shows an overall lack of difference in early stage discrimination. We find no statistically significant differences in the likelihood of making any contact when first replying to an advertisement by phone. And differences in success in setting up appointments were very small: the only significant difference from three comparisons is between white and black testers, with black testers somewhat more successful at scheduling appointments (36.1% vs. 34.8%, $p = .035$). These results provide greater confidence that our results are not driven by sample selection resulting from all three testers successfully scheduling appointments.

Finally, given the potential rarity of both encountering a set of auditors and receiving a targeted phone call from the government, there may be concerns that landlords are “spooked” by the intervention. Being spooked by the matched auditors does not threaten the internal validity of the experimental results because the design maintains measurement symmetry across all arms.³⁷ Being spooked by the targeted call from the government is not a concern because it is part of the net effect of the bundled treatment that we are interested in understanding.³⁸

Policy implications

What are the policy implications of our findings? The results on discrimination levels are relatively clear for policy makers: discrimination is a greater concern in New York

than previously believed, and this may be true elsewhere also. The evidence is especially strong for Hispanics. The implications for interventions are less clear but just as important. The challenge here is that there is suggestive evidence of strong effects, but many of the estimated effects do not hit conventional levels of significance despite the large scale of this experiment. Power issues loom large when the behavior of interest, discrimination, is relatively rare and is itself estimated with uncertainty. In such cases relying on conventional standards of significance may be overly conservative—an issue that has been of concern also in the study of rare diseases and has led to calls to alter standards in such cases (Gobburu and Pastoor 2016). Rather than altering standards, we calculate appropriate beliefs that the intervention is effective using a Bayesian framework by estimating the probability of a hypothesis (that each messaging strategy is effective) given the data. This is a quantity fundamentally distinct from the probability of the data given a hypothesis, which is the quantity of interest for determining statistical significance (Gill 1999). Understanding this alternative question is plausibly more relevant to policy makers who care about the probability that a given intervention works, rather than the probability of rejecting a specified null hypothesis. Importantly, we clarify that this is not a method to fish for statistically significant results but is instead a method to answer a substantively distinct question that can be used following future replications (given a cumulation of experimental data) to improve how policy makers form posterior beliefs about the effectiveness of different interventions.

Figure 2 plots the posterior densities along with the estimated probabilities of effectiveness (see the appendix for details about our methodology). In the bottom row, we see that, for each distribution, approximately half or less of the probability mass lies below zero: the data are not highly informative for posterior beliefs that either monitoring or punitive messages reduce discrimination against blacks. In the case of punitive messages, roughly two-thirds of probability mass lies above zero for both callbacks and offers, providing some indication that the treatment may have even increased discrimination against black testers. The top row shows a different story for Hispanics. For callbacks especially, the bulk of probability mass—at least 77%—lies beneath zero, implying posterior beliefs that the messages reduced discrimination against that group. We can be most confident in posterior beliefs about the effectiveness of the punitive treatment in reducing discrimination against Hispanics in callbacks: around 94% of the mass lies below zero, centered around the posterior mean reduction in discrimination of 6.6 points. For reference, that estimate is roughly twice the corresponding posterior mean reduction due to the monitoring treatment.

36. However, other work suggests that some discriminatory text survives (Oliveri 2010).

37. This could mean our estimates of discrimination levels in the control group are downwardly biased and could be interpreted as conservative estimates.

38. There may also be an interaction effect between encountering auditors and getting a targeted call, but we remain agnostic about that effect and recommend future research to understand it.

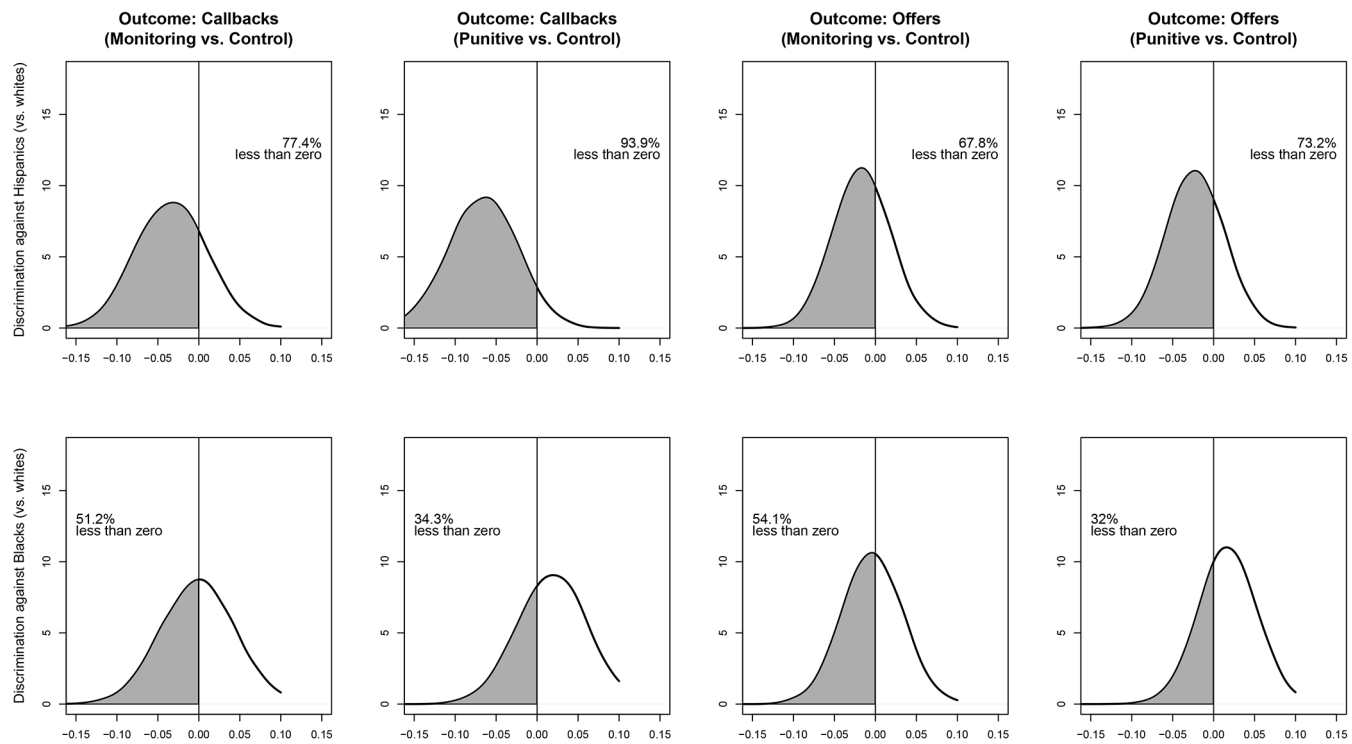


Figure 2. Posterior densities of treatment effects on discrimination against Hispanics (*top*) and African Americans (*bottom*) relative to whites. Columns correspond to combinations of outcome measure (callbacks, offers) and treatment message (punitive, monitoring) relative to a control. The area under the curve below zero is shaded in each plot, and the mean of each posterior density is shown with a dotted line. Each posterior is estimated via 10,000 Monte Carlo draws from the marginal distribution of β_1 conditional on σ^2 and y given improper uniform priors, $\beta_1 | \sigma^2, y \sim N(\hat{\beta}_1, V_{\beta_1} \sigma^2)$.

Directions for future research

Finally, we outline several directions for future research. First, an important limitation of the current field experiment is that despite the unusually large scale of our study (in comparison to prior in-person rental market audits), there is not enough statistical power in the design to detect treatment effects on the order of about 6 points. This is in part because the control group mean discrimination level is relatively low and because estimating treatment effects involves, in effect, estimating an interaction between tester race and treatment. As this is the first field experiment of its kind, it was not possible to conduct *ex ante* power analyses using previously reported effect sizes as benchmarks.³⁹ Additional experimental replications with larger sample sizes are needed to precisely estimate discrimination levels and messaging effects.

Second, additional research is needed to understand whether differential effects for Hispanic and black housing

seekers exist.⁴⁰ Existing work on the politics of policy enforcement and compliance does not explicitly develop explanations for why attempts to induce compliance with antidiscrimination law might vary by the group membership of the target of discrimination. Future work should develop and test hypotheses about the conditions under which there exists variation in baseline discrimination levels and heterogeneous effects of governmental appeals by tester and landlord group membership.⁴¹

Third, the treatments we tested in this study were, given practical considerations and constraints, bundled. They also did not test the full range of potential strategies government could deploy in preemptive messaging campaigns to reduce discrimination. Future studies could more finely operationalize treatments to test specific parameters in the decision-theoretic model that build on existing theoretical frameworks from literatures on prejudice reduction (Paluck and

39. Instead, the study was designed with the reasonable *ex ante* expectation that it was adequately powered to detect and reduce racial discrimination levels reported in rental market audit studies published before the start of the study. See the appendix for details on studies informing our priors.

40. Given space constraints, we speculate on possible explanations for mixed findings in the appendix.

41. See the appendix for an exploratory analysis of heterogeneous messaging effects by the perceived race of landlords, using the data from this experiment.

Green 2009) and on behavioral policy compliance (Weaver 2014, 2015). Future studies could also test whether effects vary by the mode of treatment delivery or by the messenger of the appeal.

Finally, replication in other housing markets and political jurisdictions would be valuable to understand how messaging effects vary across political and economic contexts. There may be different baseline discrimination levels and reasons why market actors discriminate that moderate their behavioral response to governmental appeals. There may also be different expectations among market actors about the credibility of government efforts to enforce antidiscrimination laws. How citizens perceive the capacity of the state to enforce compliance (Weaver 2014) and the degree of government forbearance in the domain of civil rights (Holland 2016)—which are partly a function of the incentives and preferences of bureaucrats and their principals (Einstein and Glick 2016; Mummolo 2018; White, Nathan, and Faller 2015)—may affect how they perceive the credibility of government signals to monitor market behavior and enforce the law, the expected intensity of government efforts to do so, and ultimately the costs of violating the law.

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