

Shedding Light on Bias: Consumer Complaint Disclosure and Racial Equity in Financial Services ^{*}

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Abstract

This study investigates how the public disclosure of consumer complaint narratives in 2015 by the Consumer Financial Protection Bureau affects racial disparities in financial services. Using triple-difference estimation, we show that minority consumers receive better treatment from financial institutions under CFPB supervision after the disclosure. These improvements manifest as higher deposit rates and lower fees in savings markets, and reduced rates for auto loans and credit cards in lending markets. Financial institutions receiving discriminatory complaints face deposit outflows. Our evidence demonstrates the broad impact of service quality disclosure in reducing racial inequalities across both savings and lending markets.

Keywords: Disclosure; Racial disparities; Financial services

JEL: G28; J15; D14

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1 Introduction

Disclosure is a widely adopted regulatory tool across various industries. Unlike traditional enforcement that relies on mandates and penalties, disclosure aims to influence stakeholders by improving transparency and providing information for informed decision-making. Studies have documented its effectiveness in various settings, from improving restaurant hygiene ([Jin and Leslie \(2003\)](#)) to enhancing environmental governance through air pollution information disclosure ([Tomar \(2023\)](#); [Barwick et al. \(2024\)](#)).

In financial services, disclosure assumes even greater importance given the sector’s vital role in shaping economic activities and daily financial decisions. While extensive research has examined the disclosure of financial institutions’ risk exposures, capital adequacy, and financial reporting ([Goldstein and Sapra \(2014\)](#); [Acharya and Ryan \(2016\)](#)), little attention has been paid to the disclosure of service quality, particularly regarding potential discrimination in day-to-day financial services. This lack of transparency makes it difficult to monitor and address persistent racial disparities in financial services.

This study investigates how disclosing financial service quality affects racial disparities across financial markets. Racial bias remains a persistent issue in financial services ([Cohen-Cole \(2011\)](#); [Faber and Friedline \(2020\)](#); [Butler et al. \(2023\)](#)) despite governmental efforts to promote fairness ([Begley and Purnanandam \(2021\)](#)). As a result of this pervasive bias, minority consumers currently receive lower-quality services compared to white consumers. Compounding this problem, limited financial literacy makes minority consumers even more vulnerable to exploitation in their interactions with financial institutions ([Guiso et al. \(2022\)](#); [Scott et al. \(2024\)](#)), ultimately leading to their exclusion from mainstream financial systems. While finding effective policy solutions is essential but challenging, our study provides the first examination of how disclosure policies impact racial discrimination in financial services, with a focus on savings and lending markets.

Our analysis demonstrates that the disclosure policy significantly reduces racial disparities across different financial services. The consistent findings in both savings and lending markets indicate disclosure’s broad effectiveness as a regulatory tool in addressing unequal treatment. In the savings market, we find that branches of financial institutions with publicly disclosed complaint narratives adjust their practices in minority communities by offering more favorable deposit rates and lower account maintenance fees after the disclosure pol-

icy. In the lending market, these branches reduce their loan rates for both auto loans and credit cards compared to the branches not under CFPB oversight. Furthermore, we find that branches of financial institutions frequently cited for discriminatory practices in complaint narratives face the threat of deposit outflows, which may serve as a key channel for the policy’s impact. Finally, we find these improvements translate into meaningful real effects for minority households. These effects include increased bank account ownership, higher auto ownership rates among low-income households, and greater confidence in credit card approval, all of which suggest improved access to mainstream financial services.

We utilize the natural experiment provided by the Consumer Financial Protection Bureau’s (CFPB) complaint narrative disclosure policy. As the primary regulatory agency for consumer financial markets in the United States, the CFPB plays a crucial role in protecting consumer interests ([Hayes et al. \(2021\)](#)). On June 25, 2015, the CFPB began publicly disclosing detailed narratives of consumer complaints with consumer permission ([CFPB \(2015\)](#)). Through this disclosure policy, the CFPB provides consumers, media, investors, and financial institutions with direct access to negative consumer evaluations of financial services. Unlike regulatory disclosures that rely on complex legal processes with limited public involvement, submitting complaints is more accessible and encourages broader public participation. This accessibility enables general stakeholders like financial consumers to become aware of issues and adjust their engagement ([Flannery et al. \(2023\)](#)). The public disclosure of these complaint narratives may bring reputation concerns and litigation risks to the targeted institutions. Our findings show that the disclosure of complaint narratives prompts financial institutions to improve both their complaint resolution outcomes and service quality.¹ We begin by employing a difference-in-difference (DID) model at the complaint level and find that, after the narratives become public, financial institutions respond more attentively to complaints with narratives than to complaints without narratives, as reflected by lower consumer dispute rates, more timely handling, and increased likelihood of monetary relief.

Beyond these ex-post handling outcomes, we explore whether financial institutions proactively improve service quality and reduce racial gaps. To identify changes in racial gaps, we exploit the CFPB disclosure policy to construct a triple-difference (DDD) model. We begin by classifying counties as majority-minority or majority-white based on their de-

¹Based on our discussions with CFPB officials, financial institutions actively monitor and analyze consumer complaints to assess and improve the quality of their products and services.

mographic composition.² Using the CFPB’s published list of regulated institutions, we then identify financial institutions under CFPB oversight and locate their branches in the RateWatch dataset. These branches, designated as treated, are subject to the CFPB’s complaint disclosure requirements, while other branches serve as controls. Our triple-difference approach examines how racial disparities in service quality between treated and control branches evolve following the disclosure policy implementation.

Our identification strategy offers a key advantage for identification: it focuses on changes in racial disparities rather than requiring direct comparability between treated and control branches. CFPB-regulated depository institutions are typically larger (with assets above the \$10 billion regulatory threshold). We impose asset-size restrictions to enhance comparability, but some unobservable differences may still remain. By introducing county racial composition (majority-minority vs majority-white) as the third dimension in our triple-difference framework, we can examine how the disclosure policy affected racial gaps. This approach relies on the assumption that, absent the disclosure policy, racial gaps would have followed parallel trends across treated and control branches. Our empirical tests support this parallel trends assumption for racial gaps, which helps address concerns about imperfect comparability between treatment and control groups.

Our study provides a comprehensive evaluation of disclosure’s impact by examining both savings and lending markets, as CFPB oversight extends across these markets. This multi-market analysis enables us to assess the broad effectiveness of disclosure in reducing discrimination across different financial services.

We begin with the savings market, which represents a crucial entry point for consumers into the financial system and plays a vital role in wealth accumulation. Despite its importance, this market has historically received less regulatory attention regarding discrimination than the lending market, making it particularly important to understand financial inclusion and disparities in wealth accumulation.³ In the savings market, the risk-free nature of this setting minimizes the possibility that disparities in treatment by financial institutions stem

²We define minorities as all non-white racial groups. A county is classified as majority-minority if its minority population exceeds 50% of the total population.

³While major consumer protection laws such as the Equal Credit Opportunity Act, Fair Housing Act, and Community Reinvestment Act aim to ensure equal access to financial services, they primarily focus on lending activities. Regulation in the savings market mainly addresses deposit insurance and bank stability through the Federal Deposit Insurance Act and Truth in Savings Act. The CFPB’s complaint-based approach provides a new regulatory channel to address racial disparities in the savings market.

from risk-related concerns, as seen in the lending market. Instead, these disparities are more likely to inherently reflect discriminatory practices. Our analysis focuses on two key dimensions of the savings market from 2013 to 2022: deposit rates and account maintenance fees. These measures capture both the return on savings that directly affects individuals’ wealth accumulation and the barriers to financial access, providing a comprehensive view of service quality in the savings market.

Our analysis reveals significant improvements in financial service terms for minority communities following the disclosure policy. For 12-month CDs with a \$10,000 minimum balance, branches in minority communities increase their deposit rates by 6.3 basis points.⁴ This pattern extends to other mainstream deposit products, generating at least \$45.8 million in annual welfare gains for minority consumers. We also find that these branches reduce their account maintenance fees by 30.5% for savings accounts compared to branches in white communities, equivalent to a decrease of \$86 in minimum balance requirements. For checking accounts, the reduction reaches 34.3%, translating to a \$1,871 decrease in minimum balance requirements in minority communities. These reductions substantially lower the entry barriers to basic banking services for minority households.

We next investigate how disclosure affects the lending market, where financial services are both high in volume and essential to consumers’ daily lives. While prior research extensively examines mortgage lending due to regulatory oversight and data availability (Li (2023b)), we focus on two understudied but important loan types: auto loans and credit cards. Unlike mortgage lending, auto loans lack structured oversight and transparency (Butler et al. (2023)), making quality disclosure critically important for consumers. Credit cards, as the most widely used credit product, significantly influence consumers’ credit access. By examining these less-regulated market segments, we assess disclosure’s effectiveness in mitigating racial biases where traditional oversight is limited. We find that branches in minority communities subject to the disclosure policy reduce their loan rates by 10.3 basis points for auto loans and by 37.2 basis points for credit card rates compared to those in predominantly white communities.

Our comprehensive analysis of both savings and lending markets captures the full spec-

⁴Adjustments to deposit rates made by branches beyond macroeconomic policies reflect their discretionary preferences in providing deposit services. Few studies examine how disclosure policies affect such branch-level decisions and their implications for minority communities.

trum of financial inclusion and economic mobility for minority households. The joint examination is particularly important because these markets are interconnected in determining household financial outcomes: savings market efficiency affects wealth accumulation capacity while lending market access influences households’ ability to make important investments and smooth consumption. We document that the disclosure policy generates opposing movements in interest rates across these two markets: deposit rates increase while loan rates decrease for minority communities. These contrasting patterns strengthen our identification strategy by ruling out alternative explanations. For example, our findings cannot be explained by monetary policy changes, which would typically move interest rates in both markets in the same direction.

Having established the broad impact of the disclosure policy, we next investigate the mechanisms driving these improvements in financial service terms. Our analysis reveals that complaint narratives serve as an important transmission channel, particularly those highlighting racial discrimination. Leveraging GPT-4o, a state-of-the-art Large Language Model (LLM) with advanced natural language understanding capabilities, we analyze consumer complaint narratives to assess potential discrimination-related issues. These narratives may shape consumer behavior and public awareness of discriminatory practices, thereby disciplining financial institutions’ pricing decisions. We conduct two analyses to substantiate this channel. First, we examine changes in branch deposit flows, which capture market-based discipline through consumer choice and directly affect banking system stability (Egan et al. (2017)). Our results show that branches of institutions with more discriminatory complaints experience significant deposit outflows in minority communities. Second, we find that greater exposure to discriminatory complaints leads to larger improvements in product pricing for minority communities, suggesting that public scrutiny motivates institutional response.

These changes in financial service terms translate into tangible economic outcomes for minority households. Using detailed household-level data, we precisely identify racial characteristics and quantify the impact on financial inclusion. Our analysis yields three key findings that demonstrate the real effects of the disclosure policy. First, we observe a significant increase in bank account ownership among low-income minority households in areas with higher concentrations of discriminatory complaints. Second, these households show notably higher auto ownership rates following the policy implementation, indicating improved access to auto financing. Third, minority households in states with more documented dis-

crimination complaints report substantially greater confidence in credit card approval after the policy takes effect. These results collectively demonstrate that enhanced transparency through disclosure leads to measurable improvements in minority households’ participation in the mainstream financial system.

This study makes several important contributions to the literature. First, we advance our understanding of the broad effects of disclosure policies. While prior research primarily examines disclosure related to corporate governance issues (Goldstein and Sapra (2014); Acharya and Ryan (2016); Leuz and Wysocki (2016); Chen et al. (2018); Bertomeu et al. (2021)), we provide the first systematic evidence on how consumer narrative disclosures shape financial service provision. Our analysis bridges previously separate literature streams of information disclosure in finance and accounting and consumer reviews in marketing (Fennis and Stroebe (2014); Zhang and Li (2021); Varga and Albuquerque (2024)) by demonstrating how public sharing of consumer experiences can drive institutional change. Furthermore, our comprehensive examination of both savings and lending markets reveals that narrative disclosures have broad effects across different financial services, enriching our understanding of disclosure as a regulatory tool.

Second, we link service quality disclosure to issues of diversity, equity, and inclusion (DEI) in financial services. While extensive research documents unexplained racial discrimination in financial services (Cohen-Cole (2011); Faber and Friedline (2020); Bartlett et al. (2022); Butler et al. (2023)), traditional regulations often face implementation challenges and may create quantity-quality trade-offs (Begley and Purnanandam (2021)). Our study demonstrates that disclosure-based regulation can effectively complement traditional approaches in promoting equal access to financial services. Unlike conventional enforcement mechanisms that depend on regulatory monitoring and legal proceedings, the disclosure of consumer narratives creates continuous market-based pressure through public scrutiny and reputational concerns. We show that this market discipline mechanism leads to meaningful improvements in both deposit and lending terms for minority communities.

Third, we contribute to the growing literature on CFPB’s regulatory effectiveness. Prior studies present mixed evidence on CFPB oversight: while some find it reduces predatory practices and stabilizes the financial system (DeFusco et al. (2020); Fuster et al. (2021); Mazur (2022)), others argue it increases compliance costs and limits credit availability (Neugebauer and Williams (2015); DeFusco et al. (2020); Fuster et al. (2021)). We extend

this literature by showing how the CFPB’s disclosure-based approach can reduce market unfairness through market discipline rather than its direct oversight role. Moreover, while existing research has extensively examined the CFPB’s impact on mortgage lending (Li (2023b); Dou et al. (2024); Dou and Roh (2024)), we show that complaint narratives affect a broader range of financial services where traditional oversight is more limited. Our analysis of both savings products and non-mortgage credit markets reveals that disclosure-based regulation can effectively address discrimination in financial services that have historically received less regulatory scrutiny. While researchers have utilized the CFPB’s complaints database to study the financial service quality of mortgage products (Begley and Purnanandam (2021)), most CFPB complaints concern non-mortgage products (see Figure A2). At a broader level, our analysis provides the first systematic evidence of how consumer narratives shape financial institutions’ behavior across different product markets.

The remainder of the paper proceeds as follows. Section 2 provides institutional background on the CFPB’s complaint narrative disclosure policy and describes our data sources. Section 3 presents our main empirical analysis examining how disclosure affects racial disparities in both savings and lending markets. Section 4 investigates the economic mechanisms through which narrative disclosures influence financial institution behavior. Section 5 analyzes household-level outcomes to quantify the real effects of the disclosure policy on minority households. Section 6 introduces the additional analyses we conduct. Section 7 concludes.

2 Background and Data

2.1 Background of CFPB Complaint Narrative Disclosure

The Consumer Financial Protection Bureau, established under the Dodd-Frank Wall Street Reform and Consumer Protection Act of 2010, serves as the chief regulatory agency protecting consumer interests in financial markets. Beginning with accepting credit card complaints in July 2011, the CFPB subsequently expanded its scope to include bank accounts, consumer loans, mortgages, and other financial services. As a regulatory agency, it addresses a wide range of consumer issues, from major service failures to serious institutional misconduct. The CFPB maintains detailed product categories within these issues, which facilitates targeted complaint submission by consumers while providing researchers the

opportunity to analyze different products by linking complaint data with product market data.

On March 19, 2015, the CFPB finalized a policy permitting public disclosure of consumer complaint narratives about financial products and services.⁵ In the complaint process, consumers describe “what happened” and can choose to make their narratives public through a checkbox option so that other stakeholders can learn from their experience. Figure A1 demonstrates the CFPB’s online complaint submission interface.⁶

The CFPB provided a buffer period of over 90 days before implementing the disclosure policy to help companies prepare for the new system. On June 25, 2015, complaint narratives became publicly accessible through the consumer complaint database. The public database includes detailed information for each complaint: the type of financial product involved, consumer zip code, submission date, the name of the financial institution involved, and the institution’s response. Since then, the initiative has attracted substantial attention from mainstream media outlets and diverse stakeholders.⁷ Figure A2 shows complaint volumes across major categories from 2013 to 2022, demonstrating growing public engagement with the CFPB’s complaint system.

Building on this growing engagement with the complaint system, we examine whether financial institutions take complaints seriously, particularly those with publicly disclosed narratives, by analyzing complaint resolution outcomes between March 19 and December 31, 2015. On March 19, 2015, the CFPB began offering consumers the option to make their complaint narratives public when filing complaints, with the actual disclosure of these narratives starting on June 25, 2015. Table A4 and Figure A4 show significant improvements in financial institutions’ complaint resolution practices following narrative disclosure. The disclosure policy leads to a 3.9% reduction in consumer disputes, a 0.5% increase in the likelihood of timely responses from institutions, and a 2.1% increase in the likelihood of

⁵The CFPB first began publishing complaint data on March 28, 2013, but this initial disclosure only included the number of complaints filed against financial institutions, not the narrative content.

⁶Most complaints are submitted via the portal. On the website, the complaint submission follows a structured five-step process. Specifically, consumers answer the following questions: “What is this complaint about?”, “What type of problem are you having?”, “What happened?”, “What company is this complaint about?” and “Who are the people involved?”. The narrative disclosure option appears in the third step. After receiving a complaint, the CFPB transmits it to the relevant financial institution and works to obtain a response for the consumer within 15 days.

⁷For example, see “Details of Consumers’ Financial Complaints to Go Public,” March 19, 2015, The Wall Street Journal. And “Consumer Finance Complaints Hit the Internet,” June 26, 2015, MarketWatch.

monetary relief for complaints with narratives. These improvements suggest that public disclosure enhances financial institutions’ responsiveness to consumer complaints, ultimately benefiting consumers.

Furthermore, our main analysis examines changes in racial disparities in the savings and lending markets following the CFPB’s public disclosure of complaint narratives in June 2015.⁸ After the disclosure policy takes effect, there are 871,960 complaints with narratives in the markets we focus, accounting for 40.7% of all complaints. A certain proportion of these complaints relates to racial discrimination (Li (2023a)). Table A1 presents three example complaints, showing key information fields including “Date received,” “Product,” “Consumer complaint narrative,” “Company,” and “Company response.” While personal information is protected, the racial bias content is evident. For instance, one narrative describes “...The problem I have is I have an excellent credit score and Capital One raise my APR to a high 27.24%. I believe this is a discriminatory practice based on my race...”.

It is important to note that although the CFPB supervises a range of companies to ensure compliance with federal consumer financial laws, not all financial institutions fall under its oversight. For those outside CFPB supervision, the narrative disclosure policy might not apply. To conduct our analysis, we utilize the CFPB’s official list of supervised depository institutions and their affiliates to categorize financial institutions in the following subsection.

2.2 Data Sources

2.2.1 RateWatch

Our empirical analysis uses RateWatch’s branch-level data on interest rates and fees for various financial products from 2013 to 2022, focusing on deposit rates, loan rates, and fees at the weekly branch level.⁹ We aggregate these weekly observations to the quarterly level.¹⁰

⁸Our analysis uses June 25, 2015 as the disclosure policy date because complaint narratives became publicly accessible on this date. Our results remain robust when using March 19, 2015, when the CFPB finalized the policy allowing consumers to share their complaints publicly.

⁹We exclude samples prior to 2013 because the CFPB made several adjustments to its complaints database during that time, which could introduce identification biases (Haendler and Heimer (2021); Dou and Roh (2024)).

¹⁰For each product, we use its last recorded interest rate or fee within each quarter as the quarterly observation. We also examine quarterly mean values as a robustness check.

To investigate institutional responses to disclosure policies, we follow [Drechsler et al. \(2017\)](#) and examine pricing behavior in the savings and lending markets by focusing on branches that actively set their rates. Specifically, we select the following products for our analysis:

Within the savings market, we examine interest rates of widely-used deposit products, particularly focusing on 12-month certificates of deposit (CDs) with balances of \$10,000 ([Drechsler et al. \(2017\)](#); [Lin \(2020\)](#); [Dlugosz et al. \(2024\)](#)). Our analysis also covers other mainstream CDs and money market accounts (MMs). Additionally, we analyze minimum balance requirements for waiving maintenance fees in both savings and checking accounts ([Migueis et al. \(2022\)](#)).

For the lending market, our analysis covers vehicle loan rates for both short-term and long-term borrowing, represented by 36-month and 72-month terms, respectively ([Yi \(2022\)](#); [Wang \(2024\)](#)). We also examine credit card interest rates, focusing on standard and premium credit card rates ([Chava et al. \(2023\)](#)).

2.2.2 CFPB Database

We use the CFPB public consumer complaints database to examine complaint resolutions following the disclosure policy implementation. From this database, we extract all complaints filed between 2015 and 2022 and identify complaints in the savings and lending markets based on the associated product information.

Under official regulations, the CFPB maintains supervisory authority over depository financial institutions with assets exceeding \$10 billion, including banks, thrifts, credit unions, and their affiliates. We match these supervised institutions to our primary datasets using the Research Statistics Supervision Discount Identifier (RSSD) from the CFPB’s regulated institution list.¹¹ The CFPB’s oversight also extends to non-depository institutions of all sizes, including mortgage originators and servicers, payday lenders, and private student lenders, as well as major participants in other consumer financial markets as defined by CFPB rules. These institutions are also considered under the influence of disclosure regulations when they appear in our datasets.

¹¹The institution list is available at (see [link](#)). For our analysis, we categorize institutions based on the list from the March 31, 2015 edition, which is the closest date to when the CFPB finalizes the disclosure policy.

2.2.3 Others

The Summary of Deposits (SOD) dataset provides branch-level deposit amounts from FDIC-insured institutions, collected annually as of June 30 from 2013 to 2022.

The Current Population Survey (CPS) dataset offers household-level bank account ownership and various household characteristics, such as income level, highest level of education, household size, and employment category. Although the CPS dataset includes annual data from 2013 to 2022, only odd-numbered years contain information on bank account ownership.

The American Community Survey (ACS) dataset supplies household-level auto ownership and household characteristic variables, covering annual data from 2013 to 2022.

The Survey of Household Economics and Decisionmaking (SHED) dataset delivers household-level self-reported confidence in credit card approval along with household characteristic variables, encompassing annual data from 2013 to 2022.

We collect county-level demographic indicators from the Bureau of Economic Analysis and the Census Bureau, including total population, racial composition, and per capita income, spanning from 2013 to 2022. The proportion of minority groups in each county is utilized to identify minority communities.

2.3 Summary Statistics

Table 1 presents summary statistics for our key variables from 2013 to 2022. Panel A of Table 1 reports the distribution of branch-level variables. These variables encompass deposit interest rates, auto loan rates, credit card rates, and account maintenance fees from the RateWatch dataset on a year-quarter basis, along with deposit amounts from the SOD dataset on an annual basis. Panel B contains household-level indicators from three survey datasets, including bank account ownership, auto ownership, and credit card approval confidence. Panel C reports annual county-level characteristics.

In compiling these statistics, we apply a cap of 100 billion total assets, including only branches from institutions with total assets below 100 billion dollars as of the second quarter of 2015. This sample restriction ensures that institutions affected by the disclosure policy and those that are not affected remain comparable without systematic differences (Li (2023b)).¹²

¹²Larger banks, in comparison to smaller banks, often have distinct capital structures and benefit from economies of scale. These institutions also face enhanced regulatory oversight following the Dodd-Frank Act.

Additionally, to analyze racial differences in financial services pricing, we define minority communities as counties where the non-white population exceeds 50% of the total population in 2015. These communities represent 7% of the national population but contain 82.5% of the nation’s minority residents.

[Insert Table 1 about here]

Furthermore, Table A3 presents the distribution of branches of institutions and their interest rate designs in our main dataset. Column (1) focuses on CFPB-oversight institutions with total assets below 100 billion dollars, while Column (2) examines non-CFPB-oversight institutions. Panel A shows that 127 CFPB-oversight institutions operate 537 branches capable of actively setting interest rates across 310 counties. Each institution maintains 1.1 branches per county, with branches adjusting interest rates 6.86 times during our sample period. For non-CFPB-oversight institutions in Column (2), the average is 1.03 branches per county, with rate adjustments occurring around 9 times, comparable to that of CFPB-supervised institutions. Panels B and C present similar distributions for auto loan and credit card rates. Auto loan rate adjustments occur more frequently, reflecting this market’s greater responsiveness to changing conditions. In contrast, fewer branches engage in credit card rate setting, suggesting that these rates are more likely to be determined by standardized policies. Overall, after excluding large financial institutions, our sample exhibits relatively balanced interest rate variations between CFPB-oversight and non-CFPB-oversight institutions, providing sufficient rate changes to quantify the impact of the disclosure policy.

3 Main Findings on Savings and Lending Products Pricing

This section empirically examines the impact of CFPB complaint narrative disclosure on racial gaps in both savings and lending markets. In the savings market, we analyze branch-level deposit interest rates and account maintenance fees. In the lending market, we focus on auto loans and credit cards as representative products.

Specifically, bank-holding companies with assets above \$100 billion must undergo more stringent supervision and participate in regulatory stress tests under the Comprehensive Capital Analysis and Review (CCAR) framework established in 2011.

3.1 Model Specification

We assess the impact of complaint narratives disclosure on racial gaps in the pricing of the four products through a triple-difference estimation approach:

$$\begin{aligned}
Y_{i,c,t} = & \alpha + \beta_1 Treat_i \times Post_t \times Minority_c \\
& + \mathbf{X}\theta + \mu_{Branch} + \mu_{YearQuarter} \\
& + \mu_{Institution \times Year} + \mu_{Product} + \epsilon_{i,c,t}
\end{aligned} \tag{1}$$

Where $Y_{i,c,t}$ represents the pricing of four prevalent financial services: deposit rates, account maintenance fees, auto loan rates, and credit card rates at branch i in county c at time t . For our treatment indicator, $Treat_i$ equals one if branch i belongs to an institution under the oversight of the CFPB in the first quarter of 2015, and zero if the branch is not under CFPB regulation. To capture the policy change timing, $Post_t$ indicates the periods before and after the CFPB made complaint narratives public, equaling one from the second quarter of 2015 onward, and zero for earlier periods. Our county classification relies on the minority population ratio, with $Minority_c$ taking a value of one for counties where the minority ratio exceeds 50%, and zero otherwise. Finally, \mathbf{X} consists of the remaining interaction terms from the triple-difference setting and county-level characteristics, including population and per capita income growth rates.

Our specification includes several fixed effects to address unobserved variations. Branch fixed effects capture branch-specific characteristics, such as service quality. Year-quarter fixed effects absorb temporal changes in financial markets, including government transitions and economic policy adjustments. Institution-year fixed effects control for pricing differences across institutions, allowing us to identify pricing differences that arise from the racial composition of served communities within the same institution.

In the estimation, apart from presenting the baseline results, we also report results with county-institution-year fixed effects to absorb time-varying county-level heterogeneity, such as institutions' local business strategies. For regressions involving multiple products, we include product fixed effects to address unobservable differences between products. We cluster standard errors at the institution, county, and year levels to handle potential correlation in error terms. The coefficient β_1 identifies how the racial gap in the pricing of the four financial services changes in the treatment group relative to the control group after the complaint

narrative disclosure. Figure A3 provides an illustration of the triple-difference model for intuitive understanding.

3.2 Results in Deposit Interest Rates

Prior regulatory and academic efforts addressing racial disparities have concentrated on the lending market. However, it is difficult to distinguish whether racial gaps in the lending market stem from discrimination or from the higher risks typically associated with minority borrowers. Individual risk levels depend on multiple factors, making it challenging to conduct empirical analyses that adequately control for all these factors. The savings market, by contrast, operates in a risk-free environment, offering a cleaner setting for analysis.

A descriptive analysis of the savings market reveals significant racial disparities, even in this risk-free setting. From 2013 to the first quarter of 2015, just before the public disclosure of complaint narratives, depositors in white-majority communities received an average deposit rate of 0.27%, while those in minority communities faced a rate of 0.17%.¹³ This difference is statistically significant at the 1% level, confirming lower deposit rates in minority communities.

We next estimate the effect of complaint narratives disclosure on deposit rates by examining the most commonly used products, with a specific focus on 12-month CDs with an account balance of \$10,000 (12MCD10K). Column (1) of Table 2 shows that the coefficient for the triple-difference term, $Treat \times Post \times Minority$, is significantly positive. Specifically, the disclosure policy increases deposit rates by 6.3 basis points in minority communities on average, equivalent to 12.1% of the sample mean. This finding indicates that treated banks narrow the racial gaps in deposit rates across communities following the disclosure policy implementation. In Column (2), we further control for unobservable confounders by including county-institution-year fixed effects. The results remain robust, with the triple interaction coefficient staying positive and significant at the 1% level.

Columns (3) and (4) extend our analysis by including data from eight types of mainstream CDs and MMs from the savings market, following the specifications of Columns (1)

¹³We calculate the weighted average deposit rates across different communities in the savings market during the pre-disclosure period using the product 12MCD10K. To enhance the comparability of descriptive results, the calculation is based on institutions with branches in both minority and white communities, with deposit amounts at the branch level used as weights.

and (2) respectively.¹⁴ To address unobservable product-level heterogeneity, we include product fixed effects in our regressions and find that the results remain positive and significant. This larger sample analysis provides strong evidence that the disclosure policy directly increases deposit rates in minority communities. The coefficient on $Treat \times Post \times Minority$ in Column (3) indicates that treated branches in minority communities raise their deposit rates by 3.9 basis points on average. We estimate that the CFPB disclosure generates economic benefits of approximately \$45.8 million annually for consumers in minority communities through higher deposit interest rates.¹⁵ Considering that this disclosure policy aims to improve banking service quality rather than specifically address racial disparities, its effect is substantial. Compared to other policies, it involves lower implementation costs and has fewer negative impacts. For instance, the Community Reinvestment Act (CRA) has been shown to have adverse effects on service quality in minority communities (Begley and Purnanandam (2021)).

[Insert Table 2 about here]

A potential limitation of our analysis is that setting a \$100 billion total asset cap for institutions might artificially restrict our sample and bias the results. To address this concern, we vary the cap threshold from \$20 to \$500 billion, setting a new cap at every \$20 billion interval and estimating the model at each threshold. Figure A5 shows that the coefficients remain significant across all thresholds, confirming the robustness of our results. The coefficients exhibit a decreasing pattern as the cap increases, suggesting stronger effects under smaller caps. This may be because smaller caps could alleviate the systematic differences arising from large financial institutions. With a \$20 billion cap (shown in the leftmost line), the disclosure policy increases deposit rates in minority communities by 7.5 basis points. Even with a \$500 billion cap (shown in the rightmost line), we still observe a 4.8 basis point

¹⁴The eight products refer to four types of CDs: 12MCD10K, 24MCD10K, 36MCD10K, and 12MCD100K, as well as four MMs: MM25K, MM50K, MM100K, and MM250K.

¹⁵We quantify the economic magnitude as follows. The average total deposit amount per treated branch in minority communities is \$174.37 million. Based on the coefficient on $Treat \times Post \times Minority$ in Column (3) of Table 2, which indicates a 3.9 basis points increase in deposit rates for treated branches in minority communities, each branch provides an additional \$0.07 million in annual interest payments to minority community depositors. The Summary of Deposits reports 928 treated branches operating in minority communities. We exclude noninterest-bearing deposits from our calculation because they are unaffected by changes in the disclosure policy. In the fourth quarter of 2015, FDIC-insured domestic noninterest-bearing deposits made up approximately 27.5% of total deposits. Based on these figures, we estimate the total increase in annual interest payments to minority communities to be at least \$45.8 million.

increase. Considering the balance of observations between treated and control branches, our baseline cap choice (marked by the red line) appears appropriate.

The validity of our triple-difference estimation relies on the assumption of parallel trends. This assumption requires that racial gaps between treated and control branches in both minority and white counties share parallel trends. When this assumption is met, we have grounds to believe that the triple-difference model can accurately estimate the true effects of the disclosure policy. We test this assumption using event study analysis based on Equation (1), which provides statistical validation and dynamic effect estimation. Figure 1 shows that in the quarters before the disclosure, no effects significantly different from zero are identified, and the estimated coefficients for the pre-trend are stable over time, showing no abnormal trends. Following the disclosure, we observe immediate and persistent changes in deposit rates for minority communities. These sustained effects are consistent with expectations, as the disclosure policy is a long-term design that has been in effect since its inception. This means that the CFPB continuously receives and discloses narratives related to complaints.

[Insert Figure 1 about here]

Additionally, we conduct several robustness tests to verify our findings. First, we examine two alternative forms of the dependent variable. One is the quarterly average interest rate of a branch, computed as the mean of its weekly interest rates within a quarter. The other one is the rate spread, defined as the difference between the rate in the last week of the quarter and the Federal Funds target rate.¹⁶ Table A5 shows that our findings are robust and not sensitive to the form of the outcome variable. Second, we replace our binary minority community indicator with a continuous minority ratio for each county. As shown in Table A6, the results remain consistent. Third, we further examine the impacts of the disclosure policy by separating minority communities into Black and non-Black minority communities. Table A7 indicates that the disclosure of complaints primarily improves deposit rates for Black communities.

Fourth, we employ a difference-in-discontinuities (diff-in-disc) design (Grembi et al. (2016); Bannedsen et al. (2022)), which measures the difference between the pre-disclosure and post-disclosure discontinuities at the \$10 billion total assets threshold. By focusing on the effect at this point, this method further addresses concerns that the control institutions

¹⁶We collect data from the Federal Reserve Economic Data (FRED) on the Fed funds target rate from 2013 to 2022 to calculate the rate spreads for our products of interest.

may not serve as an appropriate counterfactual for treated institutions.¹⁷ Our analysis in Table A8 shows that after the disclosure policy implementation, institutions above the threshold significantly increased deposit rates in minority communities, while remaining stable in white communities throughout the sample period.

Finally, branch ownership changes during the sample period could influence the validity of the results. For instance, branches in the control group may be acquired by institutions in the treated group, violating the triple-difference identification assumption. After excluding branches that experienced ownership transitions, our regression results remain robust, as shown in Table A9.

3.3 Results in Account Maintenance Fees

Despite limited research attention, account fees significantly impact consumers' financial well-being, affecting both their initial account opening decisions and long-term banking relationships. Prior qualitative studies suggest that high fees may deter consumers from participating in traditional banking (Servon (2017)). Faber and Friedline (2020) empirically shows that Black and Latinx populations face higher minimum balance requirements to avoid service fees. We investigate how the CFPB's disclosure policy improves the unfair treatment of minority consumers regarding fees, focusing on maintenance fees for checking and savings accounts, the two most prevalent account types.

We estimate the impact on maintenance fees using the same specification in Equation (1). However, the dependent variable is now maintenance fees, specifically the minimum balances required to avoid fees.

[Insert Table 3 about here]

Table 3 examines how CFPB disclosure affects racial gaps in minimum balance requirements for avoiding fees on savings and checking accounts. Column (1) shows that following the disclosure of complaint narratives, treated branches in minority communities reduce minimum balance requirements for savings accounts by 30.5% compared to white communities. Column (2) adds county-institution-year fixed effects to control potential unobservable factors, and the results remain robust. In Columns (3) and (4), we extend this analysis to

¹⁷According to regulation rules, the CFPB supervises banks, thrifts, and credit unions with assets exceeding \$10 billion and their affiliates. To ensure a sharp discontinuity at the \$10 billion asset threshold, we focus on banks, thrifts, and credit unions while excluding their affiliates.

checking accounts and find similar outcomes. Specifically, Column (3) shows a 34.3% reduction in checking account minimum balance requirements for treated branches in minority communities compared to white communities.

Based on estimates from Columns (1) and (3), this reduction translates to an \$86 decrease in minimum balance requirements for a savings account and a \$1,871 decrease for a checking account in minority communities.¹⁸ These effects are economically significant given the widespread use of these accounts. Furthermore, our event study analysis in Figure 2 shows no pre-trend concerns and demonstrates persistent effects following the policy change.

[Insert Figure 2 about here]

3.4 Results in Auto Loan Rates

Auto loans are one of the most prevalent consumer credit products in the market, yet they face considerably less regulatory oversight compared to mortgages. While mortgage lenders operate under stringent supervision and must report detailed borrower characteristics, auto lenders are subject to more limited regulatory requirements and disclosure obligations. This regulatory framework creates challenges for both monitoring potential racial disparities and conducting academic research on lending inequalities in this market. While Butler et al. (2023) document evidence of racial discrimination in auto lending using credit bureau data, there remains limited exploration of strategies to reduce such disparities, particularly the role of transparency mechanisms. We address this gap by analyzing how the public disclosure of consumer complaints affects racial disparities in auto loan pricing.

Table 4 analyzes the impact on auto loan rates using the setting in Equation (1), with auto loan rates as the dependent variable. The study focuses on two commonly selected auto loan products: 36-month short-term loans and 72-month long-term loans. We specifically examine these loan terms in the used car market, given its substantial market share and higher utilization by minority consumers.¹⁹ To account for systematic differences across loan products with different maturities, we incorporate term fixed effects. The triple interaction coefficient on $Treat \times Post \times Minority$ reveals a 10.3 basis point narrowing of racial pricing

¹⁸In this dataset, the average minimum balances set by treated branches in minority communities to avoid service fees are \$282.8 for savings accounts and \$5,455.4 for checking accounts.

¹⁹Butler et al. (2023) report that 42% of 2017 auto loan originations involved subprime borrowers (credit scores <660), with used cars comprising roughly three-quarters of such loans. Furthermore, Cox Enterprises' 2020-2021 survey shows minority buyers chose used cars 12.5% more frequently than new cars.

gaps between minority and white communities at treated branches relative to the gaps observed at control branches. This treatment effect is equivalent to 2.5% of the baseline sample mean rate, suggesting meaningful consumer welfare improvements.

Column (2) introduces county-institution-year fixed effects to absorb any unobserved heterogeneity that varies at the county-institution level over time. This stringent set of fixed effects controls for time-varying institutional characteristics within each county, such as local business strategies, relationships with local borrowers, market-specific lending policies, and the competitive dynamics in local banking markets. This specification yields a 6.8 basis point decline through the triple interaction term, equivalent to 1.7% of the sample mean. Both statistically and economically significant, these findings confirm that disclosure policies enhance credit access for minority borrowers.

[Insert Table 4 about here]

The event study analysis in Figure 3 supports our triple-difference estimation. Using the specification in Column (1) of Table 4, Figure 3 shows that prior to disclosure, racial gaps between treated and control branches are not significantly different. Policy effects remain statistically significant throughout the post-disclosure horizon, with most post-disclosure coefficients being negative. In the terminal period, minority borrowers at treated branches experience a 25 basis point rate reduction and this effect is significant at the 5% level.

[Insert Figure 3 about here]

3.5 Results in Credit Card Rates

Credit cards represent a fundamental financial product that provides households with crucial access to short-term liquidity and financial flexibility. Beyond facilitating everyday transactions, credit cards serve as an essential buffer against unexpected expenses and financial shocks. Despite its importance, the credit card market exhibits persistent racial disparities in minority consumers' access to credit cards and the interest rates they receive. This evidence underscores the need to improve the treatment of disadvantaged groups (Cohen-Cole (2011); Dwyer (2018)). In this section, we provide a potential solution by analyzing how complaint narrative disclosure policies influence institutional credit card interest rate pricing in minority communities.

To thoroughly examine the impact of the disclosure policy, we estimate interest rates

for both standard and premium credit cards, using product fixed effects to control differences in product characteristics. Table 5 presents the results. The estimated coefficient on $Treat \times Post \times Minority$ in Column (1) indicates that, compared to control branches, treated branches offer lower loan rates for consumers in minority communities, narrowing the racial gaps in pricing relative to those in white communities by 37.2 basis points, which corresponds to 3.2% of the sample mean. When we include county-institution-year fixed effects in Column (2), the results remain negative and statistically significant at the 1% level.

[Insert Table 5 about here]

Our event study analysis in Figure 4, based on the setting in Column (1) of Table 5, provides additional support for these findings. The figure illustrates that following the disclosure policy implementation, treated branches in minority communities gradually reduce loan rates. The long-term impact, as indicated by the final-period coefficient, is more pronounced than the short-term effects. This reflects increased consumer awareness of credit card misconduct, as evidenced by the surge in complaints related to credit card issues in Figure A2.

[Insert Figure 4 about here]

4 Channel Discussion

We observe that after the disclosure of complaint narratives, minority consumers experience improvements in both savings and lending markets compared to white consumers. This effect likely follows a two-stage influence process. First, the public disclosure of complaints may harm the financial institution’s reputation, leading to reduced trust among consumers and other stakeholders. These parties may respond by altering their behavior, such as withdrawing assets from institutions perceived to have discriminatory practices, thereby signaling these institutions. Second, financial institutions, worried about the behavior of stakeholders or the negative impacts they have already suffered, may take steps to enhance consumer welfare and mitigate the negative impacts of complaint disclosures on their operations.

Given that this process is complex and challenging to identify comprehensively, we provide a preliminary analysis in this section. We review existing research on both direct and indirect mechanisms through which complaint disclosures become effective. Furthermore,

we empirically examine behavioral changes in response to the disclosure policy for two key groups: consumers (as complaint initiators) and financial institutions (as complaint respondents).

Different complaints in the CFPB database may reflect various deficiencies in financial services provided by these institutions. To address our research question, we develop a GPT-4o-based analytical framework to detect discrimination in consumer complaints (see Appendix D for methodology). Recent studies highlight that LLMs have superior performance over traditional textual analysis methods (De Kok (2025)). Models like GPT-4 by OpenAI combine human-like comprehension capabilities with the scalability and efficiency of machine learning methods, achieving exceptional accuracy in complex classification tasks (Bai et al. (2023); Lopez-Lira and Tang (2023)). This makes it ideal for analyzing our extensive complaint dataset. Using systematically designed prompts, we train GPT-4o to function as a specialized complaint analyzer that evaluates narratives against established discrimination indicators. Our analysis reveals that 3.7% of complaints with narrative content contain evidence of discriminatory practices.

4.1 Consumer Behavior

Various reports and academic studies suggest that such disclosures can indeed affect an institution’s goodwill. A notable example is the case of Wells Fargo, the fourth-largest bank in the United States, which faced another phony bank accounts scandal in 2022. This incident potentially exposed the bank to fines and penalties amounting to billions of dollars. During this exposure, the complaint narratives in the CFPB attracted media attention. Some media outlets actively questioned Wells Fargo and contacted banking regulators.²⁰

The CFPB’s disclosure of complaint narratives helps external parties monitor financial institutions’ operational status. Research by Li (2023b) shows that in the mortgage lending market, an increase in complaint narratives leads consumers to reduce loan applications to banks with severe discriminatory complaints. For publicly listed companies, the disclosure of such complaints can also lead to a decrease in cumulative abnormal returns. Moreover, other institutions may recognize the harm caused by discrimination and enter markets where heavily complained-about institutions operate. The analysis by Dou and Roh (2024) of all

²⁰For more details, see: [link](#)

complaints in the mortgage market, regardless of their nature, shows consistent results. [Dou et al. \(2024\)](#) reveal that institutions can learn from their peers’ operational deficiencies through complaints and adjust their expansion strategies accordingly. These findings suggest that consumers and other stakeholders make meaningful behavioral changes in response to complaints against financial institutions.

Existing research on the impacts of complaint narratives disclosure has mainly focused on the mortgage lending market. Depositors, another major stakeholder in the savings market affected by the CFPB disclosure policy, also deserve attention for their reactions. In light of this, we analyze deposit amounts across various branches to test for deposit transfers induced by the CFPB disclosure. This analysis examines how depositors react to the CFPB disclosure policy, completing the picture of how complaint narrative disclosure affects financial market stakeholders.

[Chen et al. \(2022\)](#) show that bank information transparency can influence depositors’ behavior in transferring deposits. While the CFPB’s disclosure policy does not directly reveal financial institutions’ performance information, it exposes the public to consumers’ (negative) evaluations of these institutions. If treated branches in minority communities exhibit inappropriate practices and are exposed to complaints narratives, depositors in these branches might transfer some of their deposits after disclosure, resulting in branch-level deposit outflows.

We use changes in branch-by-year level deposit amounts from the SOD dataset to identify the impact of the disclosure policy. We modify Equation (1) by changing the time level to yearly and splitting the treatment group branches into two subgroups based on the ratio of complaints to total assets in 2015. The first group consists of branches of institutions with discriminatory complaints as a proportion of total assets exceeding the median (*HighDiscCompl*), while the second group includes branches of institutions below the median (*LowDiscCompl*). This segmentation allows us to observe whether discriminatory complaints influence deposit transfer behavior in minority communities. This setup is equivalent to estimating two triple-difference models simultaneously in one regression. Table 6 presents the results.

[Insert Table 6 about here]

Table 6 confirms the existence of deposit outflows, with varying degrees across branches with different complaint situations. The term $HighDiscCompl \times Post \times Minority$ shows that

deposit transfers are significant in treated branches within minority communities with a high ratio of discriminatory complaints, leading to a 9.1% deposit loss, as shown in Column (1). The other segmented treatment group also shows a negative coefficient for the triple interaction term but with smaller magnitudes and no statistical significance. This suggests that consumers effectively use information from complaint narratives and confirms that the public notices and utilizes the CFPB’s narrative disclosure policy. Column (2) adds county-year fixed effects to absorb unobservable disturbances, with results remaining robust.

[Insert Figure 5 about here]

We present the event study trend for branches with high and low discriminatory complaints, respectively, in Figure 5. Pre-trend does not exist in either of the two panels. Panel (a) indicates that branches with a high proportion of discriminatory complaints experience significant deposit outflows in the first year following the shock, after which the outflows gradually subside. This is consistent with our expectations, as the effects of the discriminatory complaints in 2015 are unlikely to persist over the long term. The attenuation may be attributed to adjustments made by the focal institution in response to the complaints or other changes that occurred in the following periods. Panel (b) represents the results for branches with low discriminatory complaints, where the coefficients are also negative, but none of them are statistically significant.

4.2 Heterogeneous Effects of Discrimination Complaints

Financial institutions, as respondents to complaints, have a motive to improve their management practices to mitigate potential losses from both the direct and indirect effects of complaint dissemination. Our analysis confirms that after public disclosure, complaints with narratives are resolved faster, receive more monetary relief, and result in fewer subsequent disputes. [Li \(2023b\)](#) analyzes officer career path data in the mortgage market and finds that after public disclosure, officers in institutions with many discriminatory complaints are less likely to be promoted, more likely to change jobs, and often move to smaller institutions. These responses demonstrate that financial institutions pay attention to complaint narratives. This attention may stem from concerns about reputation, potential legal risks associated with discriminatory behavior, or commitment to DEI compliance. These motivations lead institutions to take steps to improve conditions for minority consumers following

the public disclosure of complaint narratives.

Although this study cannot identify all the channels through which financial institutions improve minority consumer welfare, we provide evidence based on the varying outcomes that institutions perform when institutions respond to various complaint situations. If stakeholders can detect discriminatory complaints due to public narrative disclosure, and institutions take these seriously, we should observe different effects of disclosure on financial product pricing (i.e., interest rates) across branches with varying complaint situations. Using a setup similar to Table 6, we extend the baseline triple difference model discussed in Sections 3 by dividing the treatment into two groups based on complaint situations.

[Insert Table 7 about here]

Table 7 presents our findings. Column (1) shows differences in deposit interest rates between two groups: institutions with the ratio of discriminatory complaints to total assets above the median and those below the median. The estimated coefficients indicate a gradual decrease in disclosure impact across these groups, showing a larger estimate for the group with a high ratio of discriminatory complaints compared to the group with a low ratio. Although the disclosure policy’s effect varies among institutions with different complaint situations, the consistent direction of coefficients implies that even institutions with a low ratio of discriminatory complaints may improve the fairness of their financial services.

Column (2) presents loan interest rates in the lending market, using combined data from auto loan and credit card rates (with market-institution-year fixed effects to account for unobservable differences across markets.). The results demonstrate that branches under CFPB regulation show different policy effects based on their complaint situations. The analysis in Table 7 supports the effectiveness of the disclosure policy and suggests that awareness of complaint narratives can motivate financial institutions to place greater emphasis on fairness and equality in their financial services.

5 Real Effects

Previous analyses primarily focus on the prices of financial products and services, but whether these changes translate into tangible improvements in consumers’ lives remains unclear. Additionally, earlier research often relies on county-level population distributions to differentiate between minority and white communities, which may introduce imprecision and

potential confounding effects.²¹ To address these concerns, we extend the existing studies by examining the impact of the disclosure policy on different racial households, with a particular focus on low-income and minority households. Specifically, we assess the improvements in bank account ownership, auto ownership, and credit card approval confidence.

For more details, we first explore the racial disparities in bank account ownership, assessing how the policy has affected the likelihood of owning a checking or savings account. Next, we examine the differences in auto ownership, a key factor in economic mobility, to understand how the policy impacts vehicle access across racial groups. Finally, we investigate the effect of the policy on credit card approval confidence, focusing on how it has shaped minority households' perceptions of their ability to secure credit. Together, these results offer a comprehensive view of the policy's role in addressing racial disparities in financial access.

5.1 Results on Racial Gaps in Bank Account Ownership

We utilize data from the CPS, ACS, and SHED to examine the effects of the disclosure of complaint narratives policy on financial inclusion across three key variables: bank account ownership, auto ownership, and credit card approval confidence. Firstly, we focus on bank account ownership. The possession of a checking or savings account is a fundamental aspect of financial inclusion, providing households with essential access to financial resources and services. For low-income and minority households, being *unbanked* can lead to exclusion from critical financial tools, such as savings, credit building, and affordable loans, thereby limiting economic mobility and stability. The model specification is as follows:

$$\begin{aligned} BankAccount_{i,c,t} = & \alpha + \beta_1 DiscCompl_{i,c} \times Post_t \\ & + X\theta + \mu_{State} + \mu_{Year \times SimilarHousehold} + \epsilon_{i,c,t} \end{aligned} \quad (2)$$

Here, the dependent variable, $BankAccount_{i,c,t}$, represents whether household i in county c owns a savings or checking account in year t . Our primary focus is on account ownership within low-income households in minority samples.²² We concentrate on these households because low-income families often have lower participation rates in financial services, partic-

²¹For example, areas with higher minority populations might also be poorer or face greater financial risks, which could lead to policy effects addressing economic conditions rather than racial discrimination issues.

²²A low-income household is defined as one with an income that is above the Federal Poverty Line (FPL) but below twice the FPL. For more information about FPL, see: the poverty line guidelines.

ularly in bank account ownership, which directly impacts their ability to accumulate wealth and ensure financial security (Célerier and Matray (2019)). The key independent variable of interest, $DiscCompl_c \times Post_t$, is an interaction term that captures the effect of the policy intervention in relation to the level of complaints within the community. $DiscCompl_{i,c}$ is a binary variable indicating whether there were any discriminatory complaints in county c in 2015. $Post_t$ is a post-policy implementation indicator that equals one for years 2015 and beyond, and zero otherwise. \mathbf{X} includes the remaining interaction items from the difference-in-difference framework, along with county-level control variables such as population and the growth rates of per capita income. State fixed effects μ_{State} account for unobserved spatial heterogeneity. $\mu_{Year \times SimilarHousehold}$ absorbs systematic patterns among households with comparable demographic and socioeconomic characteristics over time. This fixed effect structure helps isolate the impact of the disclosure policy by controlling temporal patterns among similar households, thereby enhancing the identification of the policy’s causal effect. We define *SimilarHousehold* using the interaction of household characteristic variables, including income level, highest level of education, household size, and employment category (employed and unemployed). Standard errors are double clustered at the county and year level to address potential correlations.

[Insert Table 8 about here]

Table 8 presents the regression results examining the impact of complaints on account ownership across different racial households. Column (1) reports the results for minority households. The estimated coefficients of the interaction term $DiscCompl \times Post$ indicate that after the implementation of the disclosure policy, the likelihood of account ownership in minority households, particularly those in low-income brackets, increases by 4.3%, corresponding to a statistically significant change of approximately 5% of the sample mean, demonstrating a positive impact of the policy on account ownership within these households.

Column (2) reports the results for the samples of white households. The coefficient on the interaction term $DiscCompl \times Post$ is statistically significant at the 10% level, indicating an increase of 1.9% in the likelihood of account ownership after the disclosure policy implementation. This suggests that the transparency policy’s market discipline effects also facilitate enhanced banking access among lower-income white households. Such effects indicate that improved disclosure regarding service quality can benefit various demographic

groups facing financial access constraints.

5.2 Results on Racial Gaps in Auto Ownership

Auto ownership is a vital factor in facilitating economic mobility and access to essential services. It plays a crucial role in determining an individual’s ability to participate in the labor market, access education, and engage in social activities. Moreover, disparities in auto ownership reflect broader socioeconomic inequalities, making it a critical area of study for understanding and addressing racial and economic disparities. In this subsection, we analyze auto ownership at the household level. Compared to county-level classifications of minority communities, a more precise identification of race at the household level provides a clearer assessment of the real effects of the disclosure policy on auto ownership, particularly among disadvantaged households.

Following Equation (2), we change the dependent variable to auto ownership, representing whether household i in county c owns a vehicle in year t . Table 9 presents the regression results examining the impact of complaints on auto ownership across different racial households. Column (1) reports the results for minority households. The estimated coefficients of the interaction term $DiscCompl \times Post$ indicate that after the implementation of the disclosure policy, the likelihood of auto ownership in minority households, particularly those in low-income brackets, increases by 1.1%, corresponding to a statistically significant change of approximately 1.3% of the sample mean, demonstrating the positive impact of the policy on auto ownership within these communities.

[Insert Table 9 about here]

In contrast, Columns (2) report the results for the samples of white households. The coefficients on $DiscCompl \times Post$ are not statistically significant, suggesting that the policy does not have a discernible impact on auto ownership among white households. This difference in effects between minority and white households highlights the policy’s role in addressing disparities in auto ownership, particularly in minority communities where access to vehicles may be more constrained.

5.3 Results on Racial Gaps in Credit Card Approval Confidence

Credit card approval confidence serves as a key measure of financial inclusion, as it indicates individuals’ perceived ability to access credit, a crucial component for achieving economic stability and advancement. For low-income and minority households, having confidence in their ability to access credit can be instrumental in building a credit history, accessing affordable loans, and achieving long-term financial goals. In line with our previous analysis setting in Equation (2), we now focus on the households’ self-reported confidence in credit card approval.²³ This variable is categorized into three levels: not confident, somewhat confident, and very confident. This allows us to assess whether the disclosure policy has effectively enhanced financial confidence among minority households. By focusing on responses from low-income households, we aim to isolate the impact of these policy interventions on a demographic that is frequently marginalized in financial markets.

[Insert Table 10 about here]

Table 10 presents the regression results examining the impact of complaints on credit confidence across different racial households. Column (1) reports the results for minority households. The estimated coefficients of the interaction term $DiscCompl \times Post$ indicate that after the implementation of the disclosure policy, the level of credit confidence in minority households, particularly those in low-income brackets, increases by 0.286 points on the ordinal confidence scale, corresponding to a statistically significant change of approximately 14.9% of the sample mean, demonstrating the positive impact of the policy on credit confidence within these households.

In contrast, Column (2) presents the results for white households. The coefficient on the interaction term $DiscCompl \times Post$ is statistically significant but negative, indicating that white households may not expect to benefit from the disclosure policy when applying for credit cards. This finding suggests that disclosing discrimination-related complaints offsets pre-existing advantages, implying that the policy mitigates the prior advantages white households held in these counties. The substantial difference between these coefficients underscores the policy’s stronger impact on credit confidence among minorities, highlighting the critical role of the policy in addressing racial disparities in financial accessibility.

²³For the credit card approval confidence analysis, only state-level data is available for discriminatory complaints.

6 Additional Analysis

This section conducts additional explorations into the role of CFPB complaint narrative disclosures. First, we specifically examine whether the disclosure of complaints provides further improvements to the quality of financial services received by Asian consumers during the COVID-19 pandemic. Second, we introduce an online ranking system that enhances transparency by publicly displaying the extent of racial bias across institutions. This system uses results derived from regression methods to support disadvantaged consumers.

6.1 Asian Hate During Covid Period

Asian Hate emerged in early 2020 and continued into 2021, driven by various factors, such as economic pressures, ingrained societal prejudices, and widespread beliefs about the COVID-19 virus’s origin. During the pandemic, persistent safety threats in daily life were evident, with verbal harassment and civil rights violations frequently reported ([Jeung et al. \(2021\)](#)). Existing research shows that Asian Hate has led to decreased efficiency and a higher degree of pessimism among Asian professionals in the financial industry ([Agarwal et al. \(2023\)](#); [Qiao et al. \(2023\)](#)). Our analysis suggests that if Asian consumers experience unfair treatment when accessing financial services, a complaint disclosure system could serve as an effective mechanism to mitigate such issues.

To identify the impact of COVID-19, we add an interaction term based on Equation (1) (see Appendix E for details). The findings are presented in Table A10. As shown in Column (1), disclosing complaint narratives significantly mitigates the Asian deposit rate gaps at CFPB-regulated institutions during the pandemic. These institutions provided more advantageous deposit rates to Asian consumers. However, this effect is less evident in other financial products, which may be attributed to the cautious attitude of Asians towards savings influencing their complaint behavior.

6.2 Online Rank

Our findings show that disclosure policies play a crucial role in enhancing minority welfare in financial services by promoting transparency and accountability. Building on this, we have developed a ranking system that highlights how financial institutions serve different

communities, with a particular focus on disparities between minority and white areas. This platform, which builds on previous studies [Li \(2023b\)](#), already includes an initial ranking of mortgage lenders. Our latest extension provides deeper insights into how branches serve diverse communities, supporting transparency and equity in financial services.²⁴ We present an example of the online website in [Figure A6](#) (see [Appendix F](#) for details).

The new functionality ranks financial branches based on four key financial outcomes: deposit rates, maintenance fees, auto loan rates, and credit card rates. By comparing branches in minority areas to those in white areas and controlling for factors like branch, time, and state-level effects, we accurately assess the impact of racial composition on financial products. This analysis aids minority consumers in avoiding disparate treatment and promotes financial inclusion by identifying and addressing disparities in the provision of financial products across different demographic regions.

7 Conclusion

This paper comprehensively investigates the impact of the CFPB’s disclosure of complaint narratives in 2015 on reducing racial disparities across multiple financial services. Using a triple-difference design, we assess the disclosure policy’s effects on the savings and lending markets. The results show that, following the disclosure, minority communities experience general welfare improvements. In the savings market, minority communities benefit from higher interest rates and lower maintenance fees. Similarly, in the auto loan and credit card markets, minority communities see lower interest rates. We then explore the factors driving the effectiveness of the disclosure policy. We find evidence of deposit outflows linked to discriminatory practices. Additionally, among low-income minority households, the disclosure leads to increased bank account ownership, auto ownership, and confidence in credit card approval. To enhance the practical relevance of this study, we provide online rankings of racial gaps across different financial institutions and products.

Our findings support the important role of quality-based information disclosure in financial services. As disclosing this negative information about service quality can lead to improvements for disadvantaged consumers, it suggests that consumers should choose to disclose narratives when filing complaints to regulators. By doing so, they can better resolve

²⁴For more details, see: [link](#)

their issues by leveraging social supervision in addition to regulatory oversight.

The findings of this study have important implications for policymakers and regulators worldwide. Approximately 93% of developed countries have regulatory entities able to acknowledge and resolve customer complaints about financial services (Li (2023b)). However, except for the CFPB in the United States, nearly no other financial regulatory agencies in different countries choose to share consumer complaint information publicly. The results of this study demonstrate that the public disclosure of customer complaints effectively alleviates discriminatory and unfair treatment of clients in the U.S. financial system. Other countries may draw insights from this disclosure policy and develop strategies for managing and disclosing consumer complaints. Implementing such measures can foster global DEI.

This study has some limitations that future research can address. First, while the analysis examines the impact of the CFPB’s disclosure policy on racial disparities in financial services, it focuses on specific sub-markets. Important products, such as student loans, have not been tested. Future research could explore the effects of information disclosure across a broader range of financial products. Second, although this study discusses and explains some channels through which information disclosure operates, it does not fully validate the entire process. Investigating the mechanisms of information disclosure is crucial for both the financial industry and academia areas. Future studies could employ more comprehensive data and methods to explore how disclosure functions in the real world.

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Figures and Tables

Figure 1. Event Study Trend for Impact of CFPB Disclosure on Racial Gaps in Deposit Interest Rates

This figure illustrates the trend of the event study on the impact of CFPB disclosure on racial gaps in deposit interest rates. It includes outcomes from five quarters before the second quarter of 2015the initiation point of the disclosure policythrough six quarters after its implementation. Samples outside this selected span are grouped into the initial and final periods. During the estimation, we use the quarter when the disclosure occurs as the baseline period. The vertical axis displays the size of the coefficients for the triple-interaction items, and the horizontal axis represents varying periods, with period zero corresponding to the disclosure. The regression specification used to generate this figure aligns with that in Column (1) of Table 2. Black circles indicate the coefficient estimates for different periods, and the black vertical lines signify the confidence intervals (set at a 5% significance level).

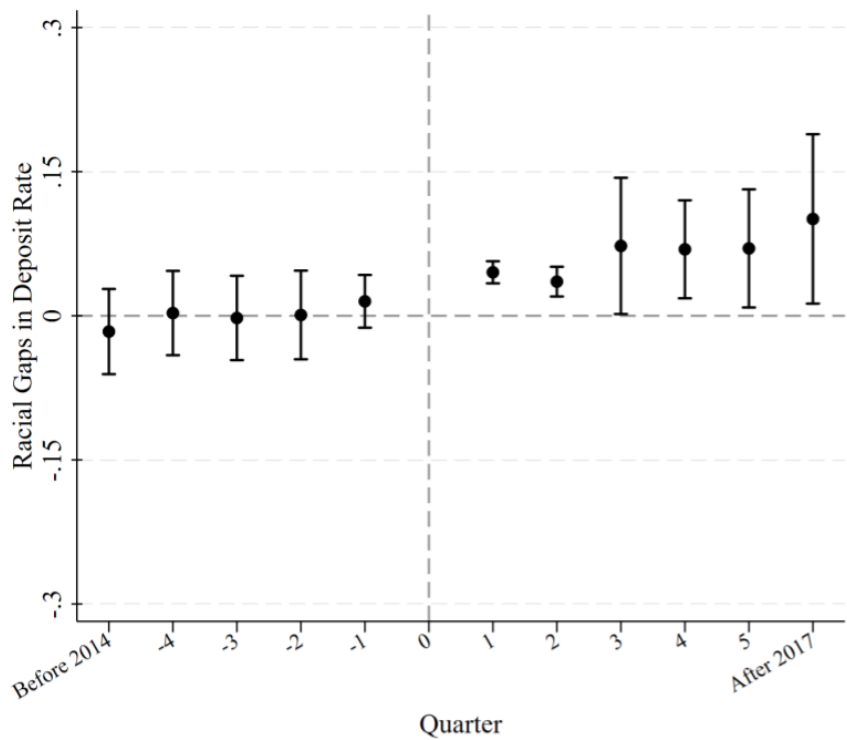


Figure 2. Event Study Trend for Impact of CFPB Disclosure on Racial Gaps in Deposit Fees

This figure illustrates the trend of the event study on the impact of CFPB disclosure on racial gaps in deposit fees. It includes outcomes from five quarters before the second quarter of 2015 the initiation point of the disclosure policy through six quarters after its implementation. Samples outside this selected span are grouped into the initial and final periods. During the estimation, we use the quarter when the disclosure occurs as the baseline period. The vertical axis displays the size of the coefficients for the triple-interaction items, and the horizontal axis represents varying periods, with period zero corresponding to the disclosure. The regression specification used to generate this figure aligns with that in Column (1) of Table 3. Black circles indicate the coefficient estimates for different periods, and the black vertical lines signify the confidence intervals (set at a 5% significance level).

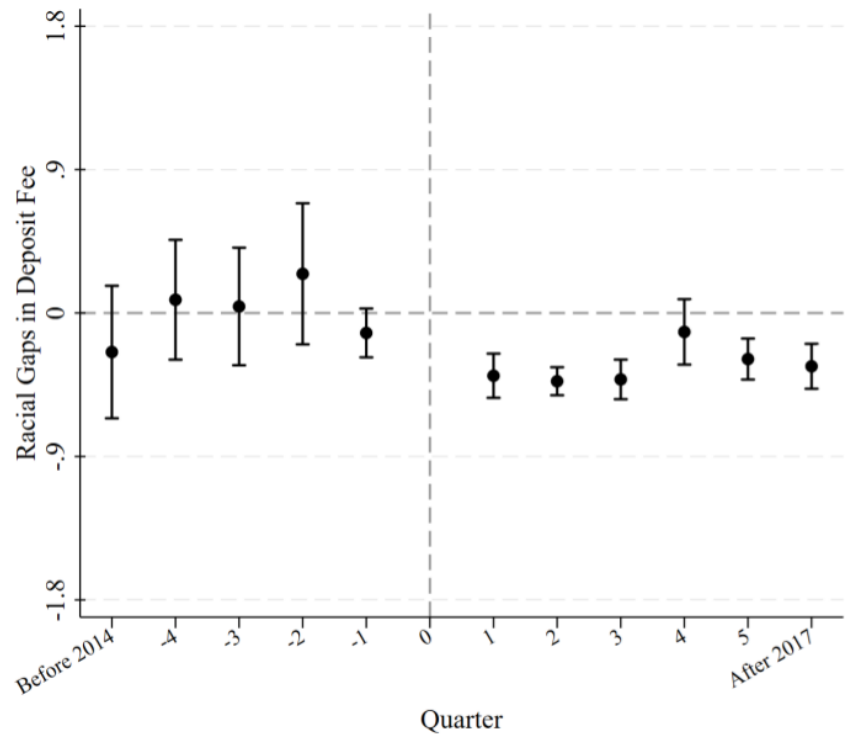


Figure 3. Event Study Trend for Impact of CFPB Disclosure on Racial Gaps in Auto Loan Rates

This figure illustrates the trend of the event study on the impact of CFPB disclosure on racial gaps in auto loan interest rates. It includes outcomes from five quarters before the second quarter of 2015 (the initiation point of the disclosure policy) through six quarters after its implementation. Samples outside this selected span are grouped into the initial and final periods. During the estimation, we use the quarter when the disclosure occurs as the baseline period. The vertical axis displays the size of the coefficients for the triple-interaction items, and the horizontal axis represents varying periods, with period zero corresponding to the disclosure. The regression specification used to generate this figure aligns with that in Column (1) of Table 4. Black circles indicate the coefficient estimates for different periods, and the black vertical lines signify the confidence intervals (set at a 5% significance level).

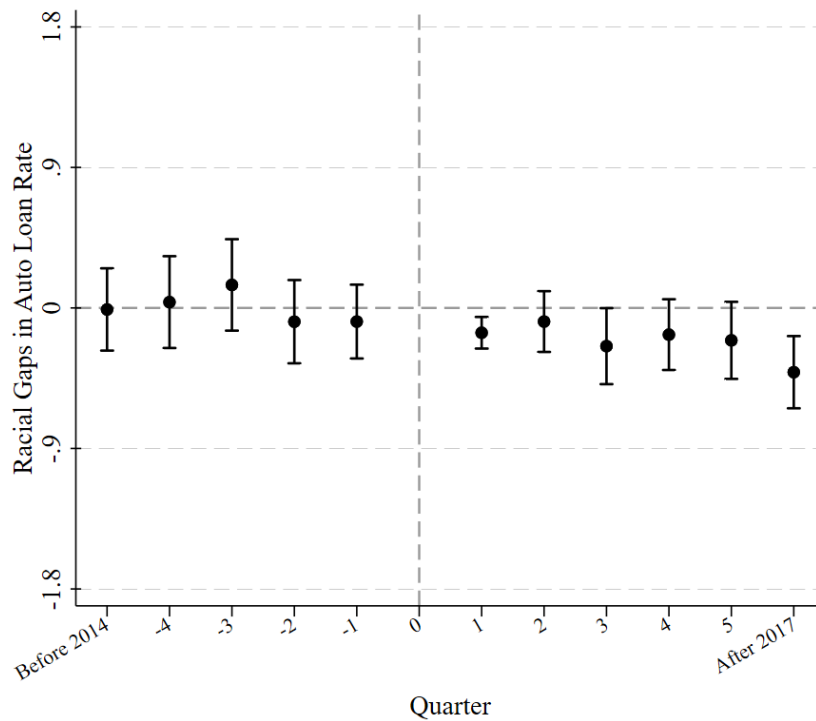


Figure 4. Event Study Trend for Impact of CFPB Disclosure on Racial Gaps in Credit Card Rates

This figure illustrates the trend of the event study on the impact of CFPB disclosure on racial gaps in credit card loan interest rates. It includes outcomes from five quarters before the second quarter of 2015 the initiation point of the disclosure policy through six quarters after its implementation. Samples outside this selected span are grouped into the initial and final periods. During the estimation, we use the quarter when the disclosure occurs as the baseline period. The vertical axis displays the size of the coefficients for the triple-interaction items, and the horizontal axis represents varying periods, with period zero corresponding to the disclosure. The regression specification used to generate this figure aligns with that in Column (1) of Table 5. Black circles indicate the coefficient estimates for different periods, and the black vertical lines signify the confidence intervals (set at a 5% significance level).

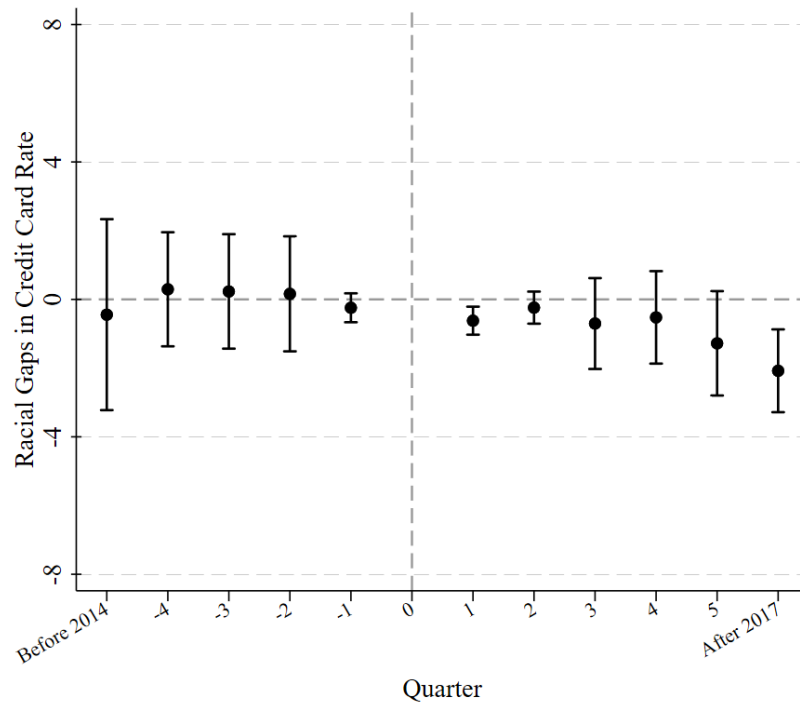


Figure 5. Event Study Trend for Impact of CFPB Disclosure on Deposit Outflows in Minority Communities Across Different Complaint Situations

This figure illustrates the trend of the event study on the impact of CFPB disclosure on deposit outflows under different complaint situations. It includes outcomes from two years before 2015 the initiation point of the disclosure policy through four years after its implementation. Samples outside this selected span are grouped into the final period. During the estimation, we use the year when the disclosure occurs as the baseline period. The vertical axis displays the size of the coefficients for the triple-interaction items, and the horizontal axis represents varying periods, with period zero corresponding to the disclosure. The regression specification used to generate this figure aligns with that in Column (1) of Table 6. Panel (a) presents the event study trend on the triple-interaction term $HighDiscCompl \times Post \times Minority$, and Panel (b) illustrates the results on $LowDiscCompl \times Post \times Minority$. Black circles indicate the coefficient estimates for different periods, and the black vertical lines signify the confidence intervals (set at a 5% significance level).

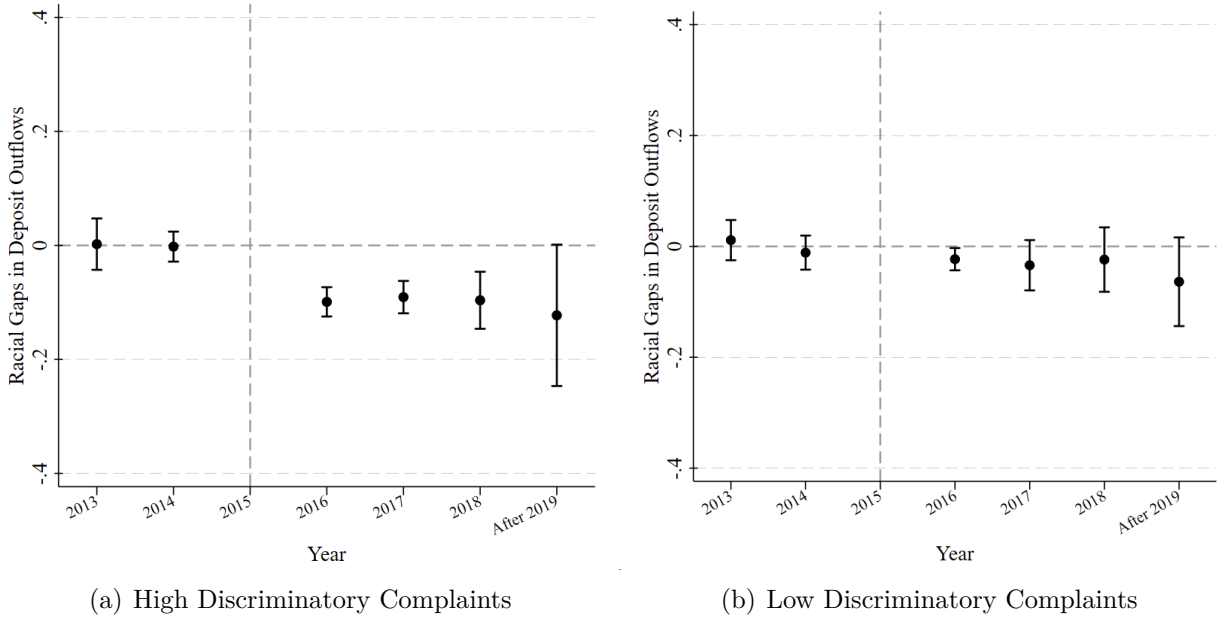


Table 1. Summary Statistics

This table shows the summary statistics for the key variables from our data sample from 2013 to 2022. Panel A highlights branch-level variables, including interest rates for deposits, auto loans, and credit cards, along with the minimum account balances necessary to bypass maintenance fees of bank accounts, sourced from RateWatch, and the deposit amounts for each branch from the Summary of Deposits (SOD) database. Within the savings market, our primary focus is on the interest rate for the product labeled “12MCD10K,” which refers to 12-month Certificates of Deposit (CDs). Appendix Table A2 presents all the CDs and Money Market (MM) products analyzed, amounting to 8 types. Regarding minimum account balances to avoid maintenance fees for savings and checking accounts. In the lending market, our discussion centers on the interest rates for auto loans and credit card loans. For auto loan rates, we focus on the rates for the most common terms, 36M and 72M. Similarly, we examine interest rates for standard card and premium card. For deposit amounts, we use the total deposits for each branch as of June 30 each year from the SOD database as a proxy. Among these five categories of indicators, deposit amounts are measured annually, while the others are quantified quarterly. Panel B highlights household-level variables measured annually, including the proportion of bank-account-owning households sourced from the Current Population Survey (CPS) database, the proportion of car-owning households sourced from the American Community Survey (ACS) database, and self-reported credit card approval confidence sourced from the Survey of Household Economics and Decisionmaking (SHED) database. Panel C summarizes county-level characteristics sourced from the Bureau of Economic Analysis (BEA) and the Census Bureau, with these data points measured annually. The variables at the county level in our analysis include the proportion of non-white populations, the total population of the county, and the growth rate of per capita income.

	(1)	(2)	(3)	(4)	(5)
	Source	N	Mean	Median	Std. Dev
<i>Panel A. Branch-Level Characteristics</i>					
Savings Market - <i>Deposit Interest Rate (%)</i>					
12MCD10K	RateWatch	310,883	0.52	0.40	0.45
Savings Market - <i>Amount to Avoid Fee (\$)</i>					
Savings Account	RateWatch	15,570	235.09	100	4023.67
Checking Account		24,680	2,777.32	1000	7949.14
Loan Market - <i>Auto Loan Interest Rate (%)</i>					
36M	RateWatch	234,346	4.12	3.75	1.64
72M		27,826	3.83	3.74	1.01
Loan Market - <i>Credit Card Loan Interest Rate (%)</i>					
Standard Credit Card Rate	RateWatch	25,370	11.98	11.99	2.57
Premium Credit Card Rate		6,518	10.57	10.25	2.08
Savings Market - <i>Deposit Amount (\$M)</i>					
Amount	SOD	521,851	99.99	42.83	1000.05

Continued on next page

Table 1 - *Continued from previous page**Panel B. Household-Level Characteristics*

Bank Account Ownership	CPS	12,159	0.92	1	0.28
Auto Ownership	ACS	1,067,572	0.85	1	0.36
Credit Card Approval Confidence	SHED	11,750	2.07	2	0.85

Panel C. County-Level Characteristics

Minority Ratio (%)	BEA	30,868	18.77	10.65	19.03
Population (<i>K</i>)	Census	30,868	104.33	25.71	333.43
Per Capita Income Growth Rate (%)	Bureau	30,868	3.82	3.58	5.65

Table 2. Impact of CFPB Disclosure on Racial Gaps in Deposit Interest Rates

This table presents the impact of CFPB disclosure on racial gaps in deposit interest rates, focusing on mainstream products. Columns (1) and (2) focus on “12MCD10K,” while Columns (3) and (4) include 8 types of CDs and MMs products. The dataset for estimating this table consists of branch-level data aggregated by year-quarter. All columns control for branch and year-quarter fixed effects, and use the annual population of each county for weighting (Célerier and Matray (2019)). Columns (1) and (3) control for demographic variables at the county level, including the logarithm of the total population and the growth rate of income per capita. These two columns also control for institution-year fixed effects. Columns (2) and (4) control for county-institution-year fixed effects instead. Through the estimation process, we obtain estimates for the key coefficient, $Treat \times Post \times Minority$, which denotes the effect of disclosure on the deposit rates in minority communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)	(3)	(4)
	Deposit Rate			
VARIABLES	12MCD10K		CDs and MMs	
$Treat \times Post \times Minority$	0.063** (0.025)	0.028*** (0.001)	0.039* (0.018)	0.028*** (0.002)
County Characteristics	YES	-	YES	-
Branch FE	YES	YES	YES	YES
Product FE	-	-	YES	YES
Year-Quarter FE	YES	YES	YES	YES
Institution-Year FE	YES	-	YES	-
County-Institution-Year FE	-	YES	-	YES
Observations	309,372	309,078	2,135,586	2,135,579
R-squared	0.879	0.883	0.735	0.738

Table 3. Impact of CFPB Disclosure on Racial Gaps in Deposit Account Fees

This table presents the impact of CFPB disclosure on racial gaps in deposit account fees, focusing on minimum balances required to avoid fees. The dataset for this analysis comprises branch-level data aggregated by year-quarter. Columns (1) and (2) are based on fees for savings accounts, while Columns (3) and (4) discuss the results for checking accounts. All columns account for county-level demographic variables, including the logarithm of the total population and the income per capita growth rate, and use the annual population of each county for weighting. All columns control for branch and year-quarter fixed effects. Columns (1) and (3) control for institution-year fixed effects, while Columns (2) and (4) control for county-institution-year fixed effects. Through the estimation process, we obtain estimates for the key coefficient, $Treat \times Post \times Minority$, which denotes the effect of disclosure on the fees in minority communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)	(3)	(4)
	Amount to Avoid Fee			
VARIABLES	Savings Account		Checking Account	
$Treat \times Post \times Minority$	-0.305*** (0.092)	-0.291*** (0.016)	-0.343** (0.146)	-0.687*** (0.024)
County Characteristics	YES	-	YES	-
Branch FE	YES	YES	YES	YES
Year-Quarter FE	YES	YES	YES	YES
Institution-Year FE	YES	-	YES	-
County-Institution-Year FE	-	YES	-	YES
Observations	8,976	8,777	10,251	10,027
R-squared	0.987	0.987	0.963	0.970

Table 4. Impact of CFPB Disclosure on Racial Gaps in Auto Loan Rates

This table presents the impact of CFPB disclosure on racial gaps in auto loan rates, focusing on auto products with terms of 36 months and 72 months. The dataset for this analysis comprises branch-level data aggregated by year-quarter. All columns include fixed effects for branch, term, and year-quarter, with the annual population of each county for weighting. Column (1) controls for county-level demographic variables, including the logarithm of total population and the growth rate of income per capita, and incorporates institution-year fixed effects. Column (2) controls for county-institution-year fixed effects instead. Through the estimation process, we obtain estimates for the key coefficient, $Treat \times Post \times Minority$, which denotes the effect of disclosure on the auto loan rates in minority communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

VARIABLES	(1)	(2)
	Auto Loan Rate	
$Treat \times Post \times Minority$	-0.103*** (0.031)	-0.068*** (0.015)
County Characteristics	YES	-
Branch FE	YES	YES
Term FE	YES	YES
Year-Quarter FE	YES	YES
Institution-Year FE	YES	-
County-Institution-Year FE	-	YES
Observations	261,902	261,878
R-squared	0.920	0.922

Table 5. Impact of CFPB Disclosure on Racial Gaps in Credit Card Rates

This table presents the impact of CFPB disclosure on racial gaps in credit card loan rates, focusing on standard and premium credit card rates. The dataset for this analysis comprises branch-level data aggregated by year-quarter. All columns control for branch, year-quarter, and product fixed effects, with the annual population of each county for weighting. Column (1) controls for county-level demographic variables, including the logarithm of the total population and the growth rate of income per capita, and incorporates institution-year fixed effects. Column (2) controls for county-institution-year fixed effects instead. Through the estimation process, we obtain estimates for the key coefficient, $Treat \times Post \times Minority$, which denotes the effect of disclosure on the credit card rates in minority communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

VARIABLES	(1)	(2)
	Credit Card Rate	
$Treat \times Post \times Minority$	-0.372*** (0.098)	-0.370*** (0.004)
County Characteristics	YES	-
Branch FE	YES	YES
Product FE	YES	YES
Year-Quarter FE	YES	YES
Institution-Year FE	YES	-
County-Institution-Year FE	-	YES
Observations	30,428	30,408
R-squared	0.916	0.918

Table 6. Impact of CFPB Disclosure on Deposit Outflows in Minority Communities
Across Different Complaint Situations

This table presents the impact of CFPB disclosure on deposit outflows in minority communities under different complaint situations. The dataset for this analysis comprises data on branch-level deposit amounts by year. In the estimation, the dependent variable is the logarithm of deposit amounts. Through the estimation process, we divide treated branches into two groups based on the complaints they received relative to their total assets in 2015. The first group includes branches of institutions with above the median numbers of discriminatory complaints as a proportion of the institution's total assets. The second group consists of branches of institutions below the median. In other words, we estimate two triple-difference models simultaneously in one regression. Columns (1) and (2) control for branch and institution-year fixed effects. Column (1) controls for demographic variables at the county level, including the logarithm of the total population and the income per capita growth rate, while Column (2) incorporates county-year fixed effects. We include all interaction terms in the triple-difference settings but only show the two core interaction terms of interest here. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

VARIABLES	(1)	(2)
	Deposit Amount	
<i>HighDiscCompl</i> × <i>Post</i> × <i>Minority</i>	-0.091*** (0.027)	-0.093** (0.031)
<i>LowDiscCompl</i> × <i>Post</i> × <i>Minority</i>	-0.035 (0.034)	-0.004 (0.034)
County Characteristics	YES	-
Branch FE	YES	YES
Institution-Year FE	YES	YES
County-Year FE	-	YES
Observations	500,952	498,800
R-squared	0.947	0.951

Table 7. Impact of CFPB Disclosure on Racial Gaps in Interest Rates Across Different Complaint Situations

This table presents the impact of CFPB disclosure on racial gaps in deposit and loan interest rates under different complaint situations. The dataset for estimating this table consists of branch-level data aggregated by year-quarter. Through the estimation process, we divided treated branches into two groups based on the complaints they received relative to their total assets in 2015. The first group includes branches of institutions with above the median numbers of discriminatory complaints as a proportion of the institution's total assets. The second group consists of branches of institutions below the median. In other words, we estimate two triple-difference models simultaneously in one regression. Column (1) is based on the baseline results from Table 2. Column (2) uses combined data from auto loan rates and credit card rates, that is, the baseline results from Table 4 and Table 5. Both columns control for demographic variables at the county level, including the logarithm of the total population and the income per capita growth rate, along with branch, year-quarter, and institution-year fixed effects. Column (2) includes product and market-year fixed effects to account for differences between the two types of loans. We include all interaction terms in the triple-difference settings but only show the two core interaction terms of interest here. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

VARIABLES	(1)	(2)
	Deposit Rate	Loan Rate
<i>HighDiscCompl</i> × <i>Post</i> × <i>Minority</i>	0.138** (0.050)	-0.399*** (0.044)
<i>LowDiscCompl</i> × <i>Post</i> × <i>Minority</i>	0.058* (0.028)	-0.094 (0.064)
County Characteristics	YES	YES
Branch FE	YES	YES
Product FE	-	YES
Year-Quarter FE	YES	YES
Institution-Year FE	YES	-
Market-Institution-Year FE	-	YES
Observations	309,372	297,838
R-squared	0.879	0.978

Table 8. Impact of CFPB Disclosure on Bank Account Ownership Across Different Racial Households

This table presents the results of the impact of discriminatory complaints on bank account ownership across different racial households, focusing on low-income households, which are defined as a household with an income above the federal poverty line but below twice the federal poverty line. The dependent variable is a binary indicator of whether a household owns a savings or checking account. Column (1) reports results for minority households and Column (2) reports results for white households. *DiscCompl* is a binary variable that equals one if a county's discriminatory complaints exceed the median and zero otherwise. The interaction term, $DiscCompl \times Post$, captures the differential effect of the policy based on the level of complaints within the community. All models control for per capita income and the logarithm of the total population at the county level. State fixed effects are included, along with the joint fixed effects of year and similar household characteristics. The similar household fixed effects encompass the interaction of income level, education level, household size, and employment category. Standard errors, shown in parentheses, are double clustered at the county and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)
	Account Ownership	
VARIABLES	Minority Households	White Households
$DiscCompl \times Post$	0.043** (0.012)	0.019* (0.008)
County Characteristics	YES	YES
State FE	YES	YES
Year-SimilarHousehold FE	YES	YES
Observations	1,791	4,615
R-squared	0.288	0.265

Table 9. Impact of CFPB Disclosure on Auto Ownership Across Different Racial Households

This table presents the results of the impact of discriminatory complaints on auto ownership across different racial households, focusing on low-income households. A low-income household has an income above the federal poverty line but below twice the federal poverty line. The dependent variable is a binary indicator of whether a household owns a vehicle. Column (1) reports results for minority households and Column (2) reports results for white households. *DiscCompl* is a binary variable that equals one if a county's discriminatory complaints exceed the median and zero otherwise. The interaction term, $DiscCompl \times Post$, captures the differential effect of the policy based on the level of complaints within the community. All models control for per capita income and the logarithm of the total population at the county level. State fixed effects are included, along with the joint fixed effects of year and similar household characteristics. The similar household fixed effects encompass the interaction of income level, education level, household size, and employment category. Standard errors, shown in parentheses, are double clustered at the county and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)
	Auto Ownership	
VARIABLES	Minority Households	White Households
$DiscCompl \times Post$	0.011*** (0.001)	-0.000 (0.001)
County Characteristics	YES	YES
State FE	YES	YES
Year-SimilarHousehold FE	YES	YES
Observations	201,600	262,354
R-squared	0.337	0.236

Table 10. Impact of CFPB Disclosure on Credit Card Approval Confidence Across Different Racial Households

This table presents the results of the impact of discriminatory complaints on credit card approval confidence across different racial households, focusing on low-income households. A low-income household has an income above the federal poverty line but below twice the federal poverty line. The dependent variable is a three-level categorical indicator representing confidence in credit card approval. Column (1) reports results for minority households and Column (2) reports results for white households. *DiscCompl* is a binary variable that equals one if a state's discriminatory complaints exceed the median and zero otherwise. The interaction term, *DiscCompl* \times *Post*, captures the differential effect of the policy based on the level of complaints within the state. All models control for per capita income and the logarithm of total population at the state level. State fixed effects are included, along with the joint fixed effects of year and similar household characteristics. The similar household fixed effects encompass the interaction of income level, education level, household size, and employment category. Standard errors, shown in parentheses, are double clustered at the state and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)
	Credit Card Approval Confidence	
VARIABLES	Minority Household	White Household
<i>DiscCompl</i> \times <i>Post</i>	0.286*** (0.076)	-0.072** (0.027)
State Characteristics	YES	YES
State FE	YES	YES
Year-SimilarHousehold FE	YES	YES
Observations	2,352	8,960
R-squared	0.228	0.114

Internet Appendix for Shedding Light on Bias: Consumer Complaint Disclosure and Racial Equity in Financial Services

Appendix A Dataset Details

Figure A1. Option to Disclose Complaint Narratives When Filing a Complaint on the CFPB Website

This figure shows the step where consumers need to describe the complaint content (“What happened”) when filing a complaint on the CFPB website. On the website, consumers need to complete five steps, answering the following questions: “What is this complaint about?”, “What type of problem are you having?”, “What happened?”, “What company is this complaint about?” and “Who are the people involved?” The choice to disclose narratives occurs in the third step, which is illustrated in this figure.

What happened?

Describe what happened, and we'll send your comments to the companies involved.

- Include dates, amounts, and actions that were taken by you or the company.
- Do not include personal information, such as your name, account number, address, Social Security number, etc. We may ask for some of this information later, to help the company identify you and your account.



☐ I want the CFPB to publish this description on consumerfinance.gov so that others can learn from my experience. The CFPB will take steps to remove my personal information from this description but someone may still be able to identify me. [Learn how it works](#). I consent to publishing this description after the CFPB has taken these steps.

Publishing this description will not affect how the CFPB handles your complaint.

Figure A2. Trend in the Number of Complaints Across Different Product Categories

This figure presents the trends in the number of complaints across different product categories from 2013 to 2022. We consider the situations for five common product types related to savings, auto loans, credit card loans, mortgage loans, and student loans (which account for the vast majority of complaints received by the CFPB). The remaining small number of complaints are classified as others. Specifically, we categorize complaints that include savings-related components from the following product classifications in the CFPB complaints database: Checking or savings account, Bank account or service, Money transfer, virtual currency or money service, Money transfers, and Virtual currency. Complaints related to auto loans are categorized under auto loans from the following classifications: Debt collection, Vehicle loan or lease, Consumer Loan, Payday loan, title loan, or personal loan, and Payday loan, title loan, personal loan, or advance loan. Complaints related to credit card services are categorized under credit card loan from the following classifications: Credit reporting, credit repair services, or other personal consumer reports, Debt collection, Credit reporting or other personal consumer reports, Credit card or prepaid card, Credit reporting, Credit card, Other financial service, and Debt or credit management. The y-axis in the graph represents the logarithmically transformed complaint amounts, while the x-axis is in months, with corresponding year labels shown. The three categories of complaints that are the focus of this study are highlighted using bold lines.

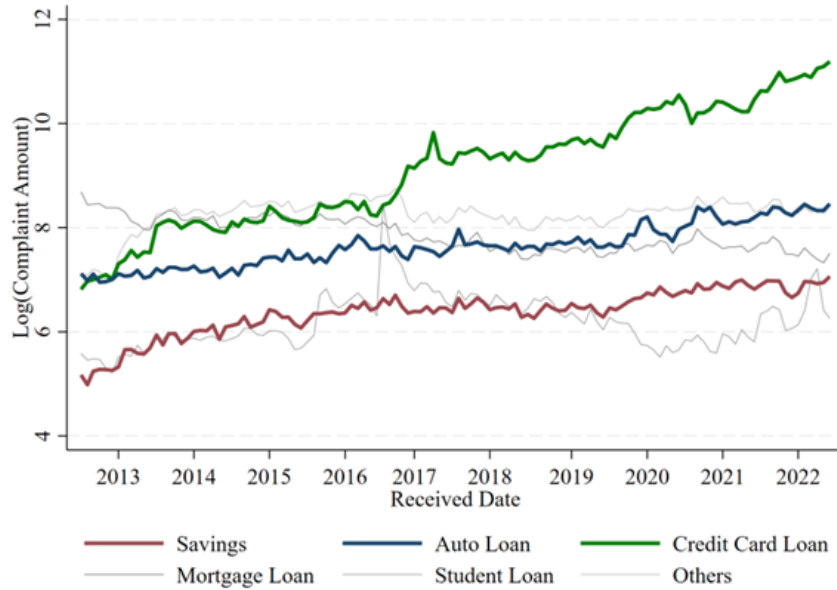


Figure A3. An Illustration of the Triple-difference Model

This figure illustrates our triple-difference identification strategy using a straightforward example. Our model differentiates the unit of analysis, branches, along two dimensions. First, branches are categorized by the regulatory status of their institutions: CFPB-oversight institutions (treatment group) versus non-oversight institutions (control group). Second, branches are classified by their geographic locations: those operating in minority communities versus those in predominantly white communities. Our triple-difference model captures how racial gaps in financial services pricing (e.g., interest rates) evolve differently for treated versus control institutions following the implementation of the CFPB disclosure policy. To provide a more intuitive understanding, we present an example in two counties in Pennsylvania. We consider two institutions, KeyBank (a CFPB-oversight institution) and WSFS Bank (an non-oversight institution), both of which operate branches in Philadelphia and Montgomery counties. Based on the minority population ratio, Philadelphia is classified as a minority community, while Montgomery is classified as a white-majority community. One key intuition behind our triple-difference estimation is to examine how the pricing at KeyBank branches in Philadelphia compares to the pricing at its branches in Montgomery, relative to the difference in pricing at WSFS branches in these two counties, and how this difference evolves after the disclosure policy is implemented.

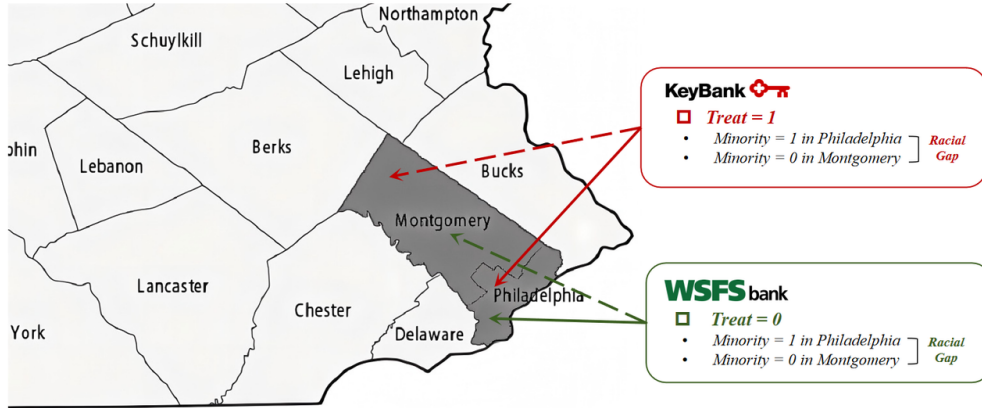


Table A1. Discriminatory Complaints Examples in Savings and Lending Markets

This table presents three examples of complaints with racially discriminatory implications from the CFPB complaints database related to the savings and lending markets. The “Consumer complaint narrative” column in the table highlights in bold the parts of the complaint narratives that indicate racial discrimination.

Date received	Details
<i>Panel A. Savings Market</i>	
Date received	12/22/22
Product	Checking or savings account
Subproduct	Checking account
Issue	Opening an account
Consumer complaint narrative	<p>On XX/XX/2022 I, XXXX XXXX, went into Amegy Bank located at XXXX XXXX XXXX XXXX, XXXXXXXX XXXX XXXX to open a business account. I was assisted by XXXX XXXX who took my documents and a grant check that was awarded to my company XXXX XXXX XXXX for the work it does for the XXXX community through the Human Rights Campaign and Showtime.</p> <p>XXXX took all my documents and stated that she needed approval to open an account. So I said okay. XXXX walked to the back and returned shortly stating that the account could not be opened because the Certificate of Formation documents were new and it would be three months before I could open the account which I found odd.</p> <p>So I left and called Amegy Bank and a representative searched and found the banking rules that said no such thing and called over to the bank while keeping me on hold. After several minutes the representative returns and explains how sorry she is for the inconvenience and hangs up.</p> <p>About twenty to thirty minutes I get a phone call from XXXX who states: I’m oversharing this but there has been a lot of business banking fraud and I was told to say that insinuating that I was there to commit fraud then tells me if I can produce documentation of where the large sum grant check came from she would gladly open the account which makes no sense because she just told me my documents were new and the reason.</p>

Continued on next page

Table A1 – Continued from previous page

	So at this point, I knew it had nothing to do with my documents. It was about me, a hardworking African-American business-woman coming into the bank with a huge check to deposit for my business that I and my business was awarded and earned and immediately being racially discriminated against. I broke down crying, disappointed and disgusted at her behavior, and reached out to XXXX, an XXXX associate who took my complaint. Currently waiting for a follow-up call.
Company	ZIONS BANCORPORATION
State	TX
Submitted via	Web
Company response to consumer	Closed with explanation
<i>Panel B. Lending Market - Auto Loan</i>	
Date received	07/15/15
Product	Consumer Loan
Subproduct	Vehicle loan
Issue	Shopping for a loan or lease
Consumer complaint narrative	I purchased a Honda Accord in 2012 at the Honda of XXXX in Ohio. I was sent an advertisement for a XXXX percent interest rate and thought I'd take advantage of the opportunity. Once I arrived at the dealership and consulted with XXXX XXXX, she informed me that not only did I not qualify for the XXXX percent interest rate but I also had the option of only XXXX car at a higher interest rate, around five percent. I thought this was odd because my credit score was XXXX at the time but with her pushy tactics, I felt forced into the deal. She told me that I did not qualify for the XXXX percent rate but there was another program I qualified for that included higher rates. Last year, I went to another dealership, Honda of XXXX in Ohio, to trade my car in and I was told about restrictions against a trade-in deal. I found out that the Honda of XXXX dealership overcharged me for the car and I was responsible for the excess charges. The salesman at the Honda of XXXX also informed me that Honda of XXXX was known for schemes such as the one I am now entangled in. With the recent lawsuit filed against Honda concerning discrimination, my situation aligns with the

Continued on next page

Table A1 – Continued from previous page

	trouble I have been and still am experiencing. It was as if Honda drew me in just to take advantage of me. I am a middle-aged black man.
Company	AMERICAN HONDA FINANCE CORP
State	OH
Submitted via	Web
Company response to consumer	Closed with explanation
<i>Panel C. Lending Market - Credit Card Loan</i>	
Date received	6/6/22
Product	Credit card or prepaid card
Subproduct	General-purpose credit card or charge card
Issue	Fees or interest
Consumer complaint narrative	On XX/XX/2022 I received an increase by Capital One without requesting it. The problem I have is I have an excellent credit score and Capital One raise my APR to a high 27.24%. I believe this is a discriminatory practice based on my race. My increase was based on my credit score and this high apr doesn't make since. Thank you
Company	CAPITAL ONE FINANCIAL CORPORATION
State	NY
Submitted via	Web
Company response to consumer	Closed with explanation

Table A2. Summary Statistics: Deposit Interest Rates of All Mainstream Deposit Products

This table presents the summary statistics for the interest rates of mainstream deposit products on a quarterly basis from 2013 to 2022. These products include four types of CDs: 12MCD10K, 24MCD10K, 36MCD10K, and 12MCD100K, as well as four MM accounts: MM25K, MM50K, MM100K, and MM250K.

	(1)	(2)	(3)	(4)	(5)	(6)
	N	Mean	Q1	Median	Q3	Std. Dev
Savings Market - Deposit Interest Rate (%)						
12MCD10K	310,883	0.52	0.25	0.40	0.60	0.45
24MCD10K	299,002	0.72	0.40	0.60	0.90	0.51
36MCD10K	282,335	0.89	0.50	0.75	1.10	0.53
12MCD100K	293,060	0.54	0.25	0.40	0.65	0.46
MM25K	282,103	0.22	0.10	0.15	0.25	0.20
MM50K	233,038	0.24	0.10	0.20	0.30	0.22
MM100K	223,580	0.28	0.13	0.20	0.35	0.25
MM250K	211,647	0.29	0.15	0.20	0.35	0.26

Table A3. Sample Composition in Interest Rates

This table shows the summary statistics for rate-setting branches belonging to CFPB-oversight or non-oversight branches in our data sample spanning from 2013 to 2022. We consider the last record of each selected product in a single quarter as the observed value for that product’s interest rate for the quarter. Panel A, B, and C highlight branch, county, and institution-level composition for deposits, auto loans, and credit cards sourced from RateWatch. Within the savings market, our primary focus is on the interest rate for the product labeled “12MCD10K.” In the lending market, our discussion centers on the interest rates for auto and credit card loans. For auto loan rates, we focus on the rates for the most common terms, 36M and 72M. We examine the standard credit card rates. Column (1) illustrates the sample composition in CFPB-oversight institutions (without large institutions), while Column (2) illustrates the composition in non-CFPB oversight institutions. The variables in our analysis include the number of institutions, the number of counties, the number of rate-setting branches, the branch-per-institution ratio, and the rates variation per branch. The branch-per-institution ratio is the average number of branches an institution sets within a county. The rates variation per branch measures the frequency of interest rate adjustments by a branch for a specific product during the sample period.

	(1)	(2)
	Oversight Institutions	Non-Oversight Institutions
<i>Panel A. Deposits</i>		
Num Institutions	127	9,176
Num Counties	310	2,498
Num Rate-Setting Branches	537	11,287
Within-County Branches Per Institution	1.1	1.03
Rates Variation Per Branch	6.86	8.65
<i>Panel B. Auto Loans</i>		
Num Institutions	62	5,190
Num Counties	156	1,808
Num Rate-Setting Branches	197	5,753
Within-County Branches Per Institution	1.05	1.02
Rates Variation Per Branch	12.61	12.02
<i>Panel C. Credit Cards</i>		
Num Institutions	49	2,229
Num Counties	81	948
Num Rate-Setting Branches	93	2,386
Within-County Branches Per Institution	1.01	1.01
Rates Variation Per Branch	4.54	2.24

Appendix B The Impact on Resolution Outcomes

In this section, we explore the impact of the CFPB’s disclosure policy on the distribution of resolution outcomes within consumer complaints. The CFPB plays a significant role in shaping financial regulation and enhancing consumer welfare, and the existence of the complaints dataset under its purview has also played a certain role in regulating the financial industry (Dou et al. (2024); Dou and Roh (2024)). Under this premise, our focus in the section is to investigate whether the CFPB’s disclosure of complaint narratives further changes the service quality of financial institutions under CFPB supervision.

Financial institutions do take the CFPB-disclosed complaint narratives seriously, which serves as the basis for our main analysis. Once the CFPB releases these narratives, their contents are likely to be noticed by various social entities, including consumers, media, investors, and possibly competing financial institutions. Concerned about potential negative reactions from these groups, the financial institutions receiving complaints are motivated to enhance their service quality more actively than before the complaint narratives were made public.

By analyzing the changes in resolution outcomes of consumer complaints, we provide a key link supporting the above path hypothesis, that is, financial institutions place more emphasis on handling complaints with narratives because the complaints narratives are made public. The existence of this phenomenon provides foundational evidence for our identification of the impact of the disclosure policy on racial gaps in financial products.

Using data from the CFPB complaints dataset, we use a difference-in-difference model to examine how complaints’ resolution outcomes change after disclosing complaint narratives to the public. We select the period from March 19, 2015, to December 31, 2015, as the analysis period. Starting on March 19, 2015, consumers could choose to make their complaint narratives public when submitting complaints to the CFPB. However, from March 19 to June 25, 2015, the CFPB did not disclose these narratives to the public to provide regulated financial institutions with some buffer time. On June 25, 2015, the CFPB began publicly sharing the complaint narratives, marking the effective initiation of the CFPB disclosure policy. This means that our pre-disclosure period is only from March 19 to June 25, 2015. During this period, consumers who submitted complaints to the CFPB faced the same situation as those who submitted complaints after June 25, 2015. This setting helps mitigate self-selection

bias from the consumers’ side, as we believe that consumers’ behavior in choosing whether to disclose complaints with narratives does not exhibit systematic differences between the pre-disclosure and after-disclosure periods.

The sample period concludes on December 31, 2015, for two main reasons. First, to avoid the potential impact of the 2016 U.S. election (Fuster et al. (2021)). Second, to balance the data structure before and after disclosure, especially given that the pre-disclosure period spans only three months. Ultimately, we retain 27,993 samples from the savings and lending markets during the study period. The regression model is as follows:

$$\begin{aligned} Y_{i,t} = & \alpha + \beta_1 Treat_i \times Post_t \\ & + X\theta + \mu_{Institution \times YearMonth} \\ & + \mu_{Zip \times YearMonth} + \mu_{Market \times YearMonth} + \epsilon_{i,t} \end{aligned}$$

Here, we divide the complaints into treatment and control groups based on whether this complaint’s narratives are disclosed to the public. $Treat_i$ equals 1 if a complaint has publicly available narratives and 0 otherwise. $Post_t$ marks the time when the CFPB provides free access to the narratives of complaints to the public, and its value equals one if the date is June 25, 2015, or later.

For the outcome variables, $Y_{i,t}$, we employ three different indicators to measure the complaints’ resolution outcomes. First, we consider the consumers’ responses after the institution responds, which is reflected by whether the consumer chooses to dispute the resolution results provided by the corresponding financial institution. Second, we examine whether the institution’s response is timely.²⁵ Third, we focus on whether complaints filed to the CFPB are resolved with monetary relief.²⁶ These three outcome variables are all in dummy form. In the regression model, β_1 is the coefficient of interest, which captures the change in resolution outcomes for complaints with narratives relative to those without narratives after the disclosure policy takes effect. \mathbf{X} incorporates the baseline terms from the difference-in-

²⁵Timely response refers to “Whether the company provided a timely response.” According to CFPB regulations, a company must provide a response within 15 calendar days. In special circumstances, a company can request up to 60 calendar days to provide a final response.

²⁶Complaints filed to the CFPB can be resolved in the following ways: closed, closed with explanation, closed with monetary relief, closed with non-monetary relief, and untimely response. Monetary relief is a resolution outcome desired by consumers, and this type of resolution has been shown to exhibit significant racial bias (Haendler and Heimer (2021)).

difference setting and county characteristics including the total population and per capita income growth rates.

Additionally, we control for a series of fixed effects to absorb the influence of unobservable factors. Institution-year-month fixed effects control for differences in institution and temporal variations, zip-year-month fixed effects capture potential influences stemming from changes in local political and economic conditions, and market-year-month fixed effects absorb heterogeneous impacts that vary over time across savings and lending markets. Standard errors are two-way clustered by institution and year-month to control for correlation within institutions and over time.

As shown in Table A4, the coefficient of $Treat \times Post$ in Column (1) is significantly positive. This indicates that, after implementing the disclosure policy, the likelihood of complaints being resolved with consumer disputed decreases by 3.9%, corresponding to 21.3% of the sample mean. Next, we estimate the impact of the disclosure policy on the efficiency of complaint responses. The coefficient for $Treat \times Post$ in Column (2) is positive and statistically significant, suggesting that narrative-containing complaints are handled more efficiently by institutions after the disclosure policy. In Column (3), this indicates that, after implementing the disclosure policy, the likelihood of complaints being resolved with monetary relief increases by 2.1%, corresponding to 47.1% of the sample mean.

These findings support that complaint resolution outcomes improve significantly after disclosing the complaint narratives to the public. Meanwhile, the estimation results of the single terms (like $Treat$) hint that merely setting an option to disclose narratives might not improve the situation for consumers who agree to make their complaint narratives public. What plays a role is actually disclosing these narratives to the public. The disclosure to the public effectively captures institutions' attention and may prompt them to be more inclined to offer favorable treatment to consumers' demands to maintain their reputation.

The premise for the difference-in-difference model to provide accurate estimates is that the parallel trend assumption can be satisfied. We conduct an event study based on the setting in the regression model to ensure the reliability of our estimation. As shown in Figure A4, we designate June 25, 2015, as the policy event date and analyze the dynamic treatment effects three months before and five months after this date.

Figure A4 (a), (b), and (c) represent the dynamic results for the three outcomes, respectively. The effects are not significantly different from zero in the months before the

disclosure, and the estimated coefficients for the pre-trend generally remain stable over time, indicating no abnormal trends. Importantly, the improvement in response outcomes occurs almost immediately at the time of disclosure. Although the improvement effects fluctuate across different outcomes, they all have long-lasting impacts, with the estimated coefficient for the fifth month showing a consistent direction to the coefficients in the initial period.

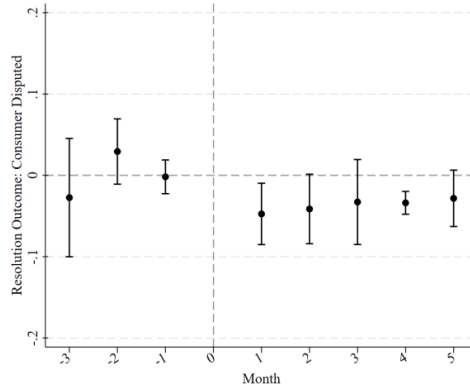
Table A4. Impact of CFPB Disclosure on Complaints' Resolution Outcomes

This table presents the impact of CFPB disclosure on complaints' resolution outcomes. The dataset for estimating this table consists of complaint-level data from March 19, 2015, to December 31, 2015. This table focuses on three resolution outcomes: Consumer Disputed is a dummy variable that takes the value of one if the consumer chooses to dispute the resolution results; Timely Response is a dummy variable that takes the value of one if the institution's response is timely; Monetary Relief is a dummy variable that takes the value of one if the complaint is resolved with monetary relief for the consumer. For the independent variables, we focus on the difference-in-difference interaction term, $Treat \times Post$, which denotes the effect of disclosure on the complaints' resolution outcomes. All columns control for institution-year-month, zip-year-month, and market-year-month fixed effects. Standard errors, shown in parentheses, are double clustered at the institution and year-month levels. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

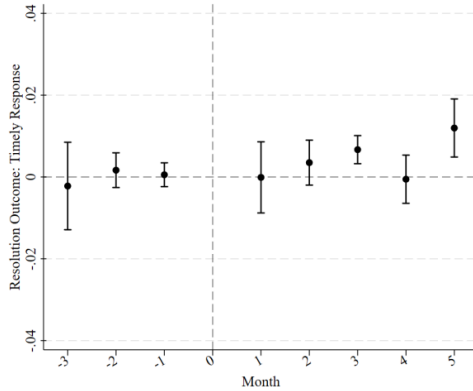
VARIABLES	(1)	(2)	(3)
	Consumer Disputed	Timely Response	Monetary Relief
$Treat \times Post$	-0.039** (0.012)	0.005* (0.002)	0.021*** (0.006)
Institution-Year-Month FE	YES	YES	YES
Zip-Year-Month FE	YES	YES	YES
Market-Year-Month FE	YES	YES	YES
Observations	27,993	27,993	27,993
R-squared	0.542	0.601	0.559

Figure A4. Event Study Trend for Impact of CFPB Disclosure
on Complaints' Resolution Outcomes

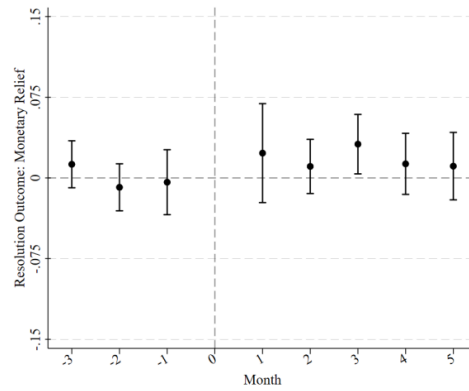
This figure illustrates the trend of an event study assessing the impact of the CFPB disclosure policy on the complaints' resolution outcomes. It addresses three resolution outcomes: In panel (a), Consumer Disputed evaluates whether the consumer opts to dispute the resolution outcomes provided by the respective financial institution. In panel (b), Timely Response assesses whether the institution's response occurs in a timely manner. In panel (c), Monetary Relief measures whether complaints filed to the CFPB result in monetary relief for the complainant. This figure covers the results from three months before the initiation of the disclosure policy on June 25, 2015, to five months after its implementation. For the fifth month after disclosure, the last few days of 2015 (i.e., December 25 to 31, 2015) are also included. During the estimation, we use the month when the disclosure occurs as the baseline period. The vertical axis displays the magnitude of the coefficients for the double-interaction terms, $Treat \times Post$ and the horizontal axis represents various periods, with period zero corresponding to the time of disclosure. Black circles indicate the coefficient estimates for different periods, and the black vertical lines signify the confidence intervals (set at a 5% significance level).



(a) Consumer Disputed



(b) Timely Response



(c) Monetary Relief

Appendix C Robustness Checks for the Impact on Financial Services

Figure A5. Impact of CFPB Disclosure on Racial Gaps in Deposit Interest Rate Across Asset Size Caps

This figure illustrates the impact of CFPB disclosure on racial gaps in deposit interest rates across different asset size caps. The horizontal axis in this figure represents the caps used for sample selection. We perform an estimation every \$20 billion from \$20 to \$500 billion. The vertical axis in this figure represents the size of the coefficients for the triple-interaction items in the triple-difference model estimated at different cap thresholds. These coefficients are estimated based on the setting in Column (1) of Table 2. The dots represent the coefficient values, and the lines extending above and below each dot represent their confidence intervals (set at a 5% significance level). The red line in this figure shows the estimated results for this paper's baseline cap, which includes institutions with total assets under \$100 billion.

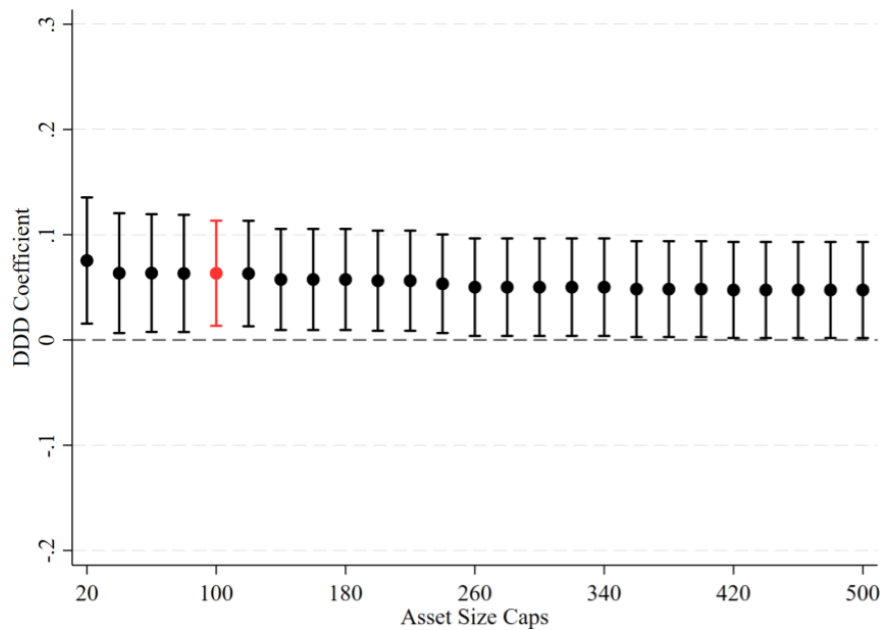


Table A5. Impact of CFPB Disclosure on Racial Gaps in Savings Market: Sensitivity with Quarterly Rate Methods

This table presents the impact of CFPB disclosure on racial gaps in rate sensitivity with quarterly rate methods. Columns (1) and (2) use the average rate of a branch within a quarter, while Columns (3) and (4) use the rate spread, calculated as the last-week rate minus the Fed funds target rate, as outcome variables. All columns control for branch and year-quarter fixed effects. Columns (1) and (3) control for demographic variables at the county level and institution-year fixed effects. Columns (2) and (4) control for county-institution-year fixed effects. Through the estimation process, we obtain estimates of the key coefficient, $Treat \times Post \times Minority$, which denotes the effect of disclosure on the deposit rates in minority communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)	(3)	(4)
	Deposit Rate			
VARIABLES	Mean Rate		Rate Spread	
$Treat \times Post \times Minority$	0.051* (0.023)	0.025*** (0.000)	0.063** (0.025)	0.028*** (0.001)
County Characteristics	YES	-	YES	-
Branch FE	YES	YES	YES	YES
Year-Quarter FE	YES	YES	YES	YES
Institution-Year FE	YES	-	YES	-
County-Institution-Year FE	-	YES	-	YES
Observations	309,372	309,078	309,372	309,078
R-squared	0.886	0.890	0.941	0.943

Table A6. Impact of CFPB Disclosure on Racial Gaps in Savings Market: Minority Ratio

This table presents the impact of CFPB disclosure on racial gaps in deposit interest rates. We replace the dummy variable *Minority* with the continuous variable *MinorityRatio*, which comes from the 2015 county-level demographic data. All columns control for branch and year-quarter fixed effects. Column (1) controls for demographic variables at the county level and institution-year fixed effects, while Column (2) controls for county-institution-year fixed effects. Through the estimation process, we obtain estimates of the key coefficient, $Treat \times Post \times MinorityRatio$, which denotes the effect of disclosure on the deposit rates in communities with different minority ratios. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

VARIABLES	(1)	(2)
	Deposit Rate	
$Treat \times Post \times MinorityRatio$	0.096** (0.036)	0.137*** (0.000)
County Characteristics	YES	-
Branch FE	YES	YES
Year-Quarter FE	YES	YES
Institution-Year FE	YES	-
County-Institution-Year FE	-	YES
Observations	309,317	309,023
R-squared	0.879	0.883

Table A7. Impact of CFPB Disclosure on Racial Gaps in Savings Market: Different Races

This table presents the results on the impact of CFPB disclosure on racial gaps in deposit interest rates across different racial groups. We modify the definition of the Minority dummy variable, creating county-level Minority variables based on the distribution (tertile value) of the proportion of Black individuals (“Black”) and the combined proportion of Hispanic and Asian individuals (“NonBlack”) in different counties. Columns (1) and (2) display the estimated effects of disclosure for each of these racial groups on “12MCD10K”. All columns control for demographic variables at the county level. We also control for branch, year-quarter, and institution-year fixed effects. Throughout the estimation process, we include all interaction terms and present the main triple interaction terms in this table. The key estimated coefficients denote the changes in deposit rates following disclosure in communities with minority ratios in the upper and middle tertiles. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)
	Deposit Rate	
VARIABLES	Black	Non-Black
<i>Treat×Post×UpperTertile</i>	0.027*** (0.006)	-0.019 (0.016)
<i>Treat×Post×MiddleTertile</i>	0.004 (0.013)	-0.017 (0.018)
County Characteristics	YES	YES
Branch FE	YES	YES
Year-Quarter FE	YES	YES
Institution-Year FE	YES	YES
Observations	311,100	311,100
R-squared	0.879	0.879

Table A8. Impact of CFPB Disclosure on Racial Gaps in Savings Market:
Difference-in-Discontinuities Model

This table presents the results on the impact of CFPB disclosure on racial gaps in deposit rates based on a difference-in-discontinuities model. We focus our analysis on banks, thrifts, and credit unions, and we have excluded data from institutions with total assets of less than \$10 billion that were subject to treatment (i.e., CFPB-regulated affiliates). This approach captures the difference in the discontinuity before and after the disclosure at the \$10 billion total assets threshold. Specifically, we estimate the following specification separately for minority and white communities:

$$\begin{aligned} DepositRate_{i,c,t} = & \alpha_i + \alpha_1(Asset_{i,t} - 10billion) + Over10_{i,t}(\beta_0 + \beta_1(Asset_{i,t} - 10billion)) \\ & + Post_t[\gamma_0 + \gamma_1(Asset_{i,t} - 10billion) + Over10_{i,t}(\delta_0 + \delta_1(Asset_{i,t} - 10billion))] \\ & + \mu_{Branch} + \epsilon_{i,c,t} \end{aligned}$$

Where $DepositRate_{i,c,t}$ represents the deposit interest rates of branch i in county c at time t across 12MCD10K. $Over10_{i,t}$ indicates whether the total assets of a financial institution exceed the \$10 billion threshold, based on the first quarter of 2015 data. $Post_t$ takes a value of one when the time t corresponds to the period after the second quarter of 2015 and zero otherwise. We include individual (branch) fixed effects α_i . The coefficient δ_0 is the diff-in-disc estimator and identifies the treatment effect of the CFPB disclosure. Columns (1) and (2) present the results for minority communities, while Columns (3) and (4) focus on white communities. Columns (1) and (3) use the full sample, while Columns (2) and (4) focus on institutions with total assets ranging from \$1 billion to \$20 billion. Standard errors, shown in parentheses, are clustered at the institution level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)	(3)	(4)
	Deposit Rate			
	Minority		White	
VARIABLES	All	[1,20]	All	[1,20]
<i>Diff-in-Disc</i>	0.037*** (0.011)	0.090*** (0.032)	0.008 (0.005)	-0.009 (0.015)
Branch FE	YES	YES	YES	YES
Observations	4,943	1,047	116,401	23,804
R-squared	0.418	0.468	0.401	0.414

Table A9. Impact of CFPB Disclosure on Racial Gaps in Savings Market: Branch Ownership Change

This table presents the results on the impact of CFPB disclosure on racial gaps in deposit interest rates considering the branch ownership changes. We excluded branch observations where ownership changed during the sample period. These exclusions are limited to branches whose institutions were altered and do not include newly established branches. All columns control for branch and year-quarter fixed effects. Column (1) controls for county-level demographic variables and institution-year fixed effects, while Column (2) controls for county-institution-year fixed effects. Through the estimation process, we obtain estimates for the key coefficient, $Treat \times Post \times Minority$, which denotes the effect of disclosure on the deposit rates in minority communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

VARIABLES	(1)	(2)
	Deposit Rate	
$Treat \times Post \times Minority$	0.055** (0.019)	0.031*** (0.001)
County Characteristics	YES	-
Branch FE	YES	YES
Year-Quarter FE	YES	YES
Institution-Year FE	YES	-
County-Institution-Year FE	-	YES
Observations	292,962	292,695
R-squared	0.879	0.884

Appendix D AI-Powered Identification of Discrimination in Complaint Narratives

Identifying discrimination patterns in consumer complaints is essential for understanding disparities in financial services, yet this issue remains underexplored. The CFPB complaint database, with its rich narrative content, provides a valuable resource for uncovering these patterns and assessing the implications of such disparate treatment.

Prior research has made initial attempts to extract relevant text from the CFPB database. For example, [Bian et al. \(2024\)](#) construct a zero-shot learning model using a word list to identify fraud-related complaints in the CFPB database. [Li \(2023a\)](#) analyzes discrimination-related topics using a word list approach. Building on these efforts, we introduce a novel analytical framework that leverages large language models (LLMs), specifically OpenAI’s advanced GPT-4o model, to detect discrimination patterns in complaint narratives. Compared to conventional methods, LLMs excel at contextual reasoning and implicit bias detection ([De Kok \(2025\)](#)), making them particularly suited for this task.

Our implementation follows [De Kok \(2025\)](#)’s best practices for LLM applications. We design GPT-4o as a specialized CFPB complaint analyst through hierarchical prompt engineering, decomposing discrimination identification into five structured components: (1) general instructions for comprehensive task understanding; (2) definition of discrimination with two key factors: customers’ feelings and their treatment comparisons with others; (3) classification standards for various forms of discrimination; (4) reasoning process guidance for systematic analysis; and (5) output format specifications for consistent results.

We employ the widely-used few-shot learning approach by including two contrasting examples in the prompt. The first example presents a negative case about a bank’s penny acceptance policy that involves no discrimination. The second example shows a positive case where a minority customer reports being denied a mortgage loan despite meeting all qualification requirements, while similar non-minority applicants receive approval. We present the negative case first to establish a baseline understanding of non-discriminatory complaints, which helps the model better distinguish actual discrimination from general service issues through learning through contrasting examples. This approach reduces the misclassification of routine service issues as discriminatory incidents, recognizing that the majority of customer grievances stem from operational and service quality concerns rather than discrim-

inatory practices.

Here is the implemented prompt:

You are an AI assistant specialized in analyzing consumer complaints for the Consumer Financial Protection Bureau (CFPB). Your task is to identify discrimination-related content in complaint narratives and determine if the complainant experienced discrimination or unfair treatment.

1. General Instructions

(1) Carefully read and analyze the above narrative.

(2) Determine if this complaint contains content related to discrimination or unfair treatment.

(3) Consider a broad definition of discrimination, including any indication of unfair or unequal treatment that the complainant perceives.

2. Key Elements of Discrimination When analyzing the complaint, pay special attention to:

(1) The complainant's feelings of discrimination or unfair treatment

(2) Any comparisons the complainant makes between their treatment and that of others

3. Forms of discrimination When analyzing the complaint, consider the following forms of discrimination or unfair treatment:

(1) Explicit or implicit discrimination based on protected characteristics (race and ethnicity)

(2) Consider the following forms of discrimination or unfair treatment:

- Approval discrimination: Higher likelihood of application rejections

- Interest rate discrimination: Being charged higher interest rates

- Service quality discrimination: Poorer service experience or staff attitudes

- Fee discrimination: Being charged higher or additional fees

- Any other form of unfair or unequal treatment perceived by the complainant

4. Reasoning Process Use your reasoning skills to analyze the narrative. Consider the following questions:

(1) Does the complaint explicitly mention discrimination or unfair treatment?

(2) Are there implicit indications of discriminatory practices or unfair treatment?

(3) Does the complaint describe treatment that aligns with any of the discrimination types mentioned above?

(4) Does the complainant express feeling unfairly treated, even if not using the word "discrimination"?

(5) Is there a comparison made between the complainant's treatment and that of others, suggesting potential disparity?

5. Output Format After your analysis, provide your output in the following JSON format:

```
{  
  "is_discrimination": "<Binary classification (0 or 1) of the complaint>",  
  "confidence_score": "<Confidence score (0.0-1.0)>",  
  "reason": "<Reason for your classification>"  
}
```

[COMPLAINT INPUT HERE]

Here are some examples:

<examples>

<example>

<COMPLAINT_NARRATIVE>

"US Bank branch refused to accept pennies in a jar, about \$10.00 dollars worth. Bank employees said no US Bank in my area accepts large amounts of pennies. I called national customer service who said each branch can have its own policy, but couldn't tell me a location that actually would accept pennies.

Apparently, the Federal Reserve doesn't have a regulation requiring banks to accept pennies. If that's the case, what value do pennies have if nobody is willing to accept them? If banks refuse pennies, and the Fed doesn't mandate banks to accept them, then obviously neither one of them wants the pennies. This makes pennies worthless, so why would I, or any consumer, want them? What if consumers refused pennies as a form of change from merchants because they have no value?

Finally, if the policy of banks and the Fed have rendered pennies worthless, why can't consumers melt them for the copper value? Pennies can't be melted because they are US currency, but shouldn't anything defined as US currency have a value of more than xxxx?

Either the Federal Reserve needs to require banks to accept pennies, or it needs to allow consumers to melt them. Thank you."

</COMPLAINT_NARRATIVE>

```

<ideal_output>
[{"is_discrimination": 0,
  "confidence_score": 0.85,
  "reason": "While the complainant expresses frustration with bank policy regarding penny
acceptance, the complaint does not indicate discriminatory treatment. The policy appears
to be applied uniformly across all customers at US Bank branches in the area. The issue
relates to general banking policy and currency acceptance rather than unfair treatment or
discrimination against any particular individual or group. The complainant's concerns focus
on the practical value and utility of pennies as currency rather than experiencing disparate
treatment."
}]

```

```

</ideal_output>
</example>
<example>
<COMPLAINT_NARRATIVE>
  I believe that I am being discriminated against because I disclosed my race as XXXX on
  Section X of the XXXX loan application. I would greatly appreciate it if this could be looked
  into to ensure that Bank of America didn't discriminate against me by showing that they
  also denied mortgage loans to people of other races, particularly XXXX people, with similar
  credit, income or debt-to-income ratio, savings, educational, and employment backgrounds
  as me.

```

```

  Quick summary of my background: I have excellent credit, my credit score is over XXXX.
  My 2 employment gaps greater than 30 days were related to school. I have a XXXX XXXX
  and currently in XXXX XXXX seeking a XXXX. I work full time as a mortgage loan advisor
  where I earn over $45000.00 annually. I have savings of $30000.00. The house I was looking
  to purchase cost $180000.00.

```

```

</COMPLAINT_NARRATIVE>

```

```

<ideal_output>

```

```

[{"is_discrimination": 1,

```

```

  "confidence_score": 0.95,

```

```

  "reason": "The complaint explicitly alleges racial discrimination in mortgage lending by

```

Bank of America. The complainant directly states 'I believe that I am being discriminated against because I disclosed my race' on the loan application. They specifically request an investigation to compare their loan denial with approvals for people of other races who have similar qualifications. The complainant provides detailed evidence of their strong qualifications (excellent credit score over XXXX, stable employment with \$45,000 annual income, \$30,000 in savings, advanced education) to support their belief that the denial was based on race rather than legitimate lending criteria. This represents a clear case of perceived discrimination in lending practices based on a protected characteristic (race), with the complainant making explicit comparisons to how others of different races might be treated in similar circumstances." }}

</ideal_output>

</example>

</examples>

The output uses a standardized JSON format with three key elements: (1) a binary classification indicating the presence of discrimination; (2) a confidence score ranging from zero to one for uncertainty quantification; and (3) a reasoning explanation for classification decisions.

We apply this framework to identify discrimination in complaint narratives across various financial services. Specifically, among the complaints from our target market in 2015, we identified 20,676 complaints containing narrative descriptions. Within these narrative-containing complaints, 774 were found to contain discriminatory content, accounting for 3.7%. Our analysis shows a slightly lower identification rate than traditional keyword-based method (Li (2023a)). This suggests that LLMs achieve more precise identification by semantically excluding content irrelevant to discrimination.

Below, we list a few example complaints with varying discrimination classification scores:

Complaint #1450743

Narrative: "I financed a vehicle thru Ally financial entering a retail installment contract on XXXX/XXXX/XXXX. The Consumer Financial Protection Bureau and U.S Department of Justice investigate Ally's auto financing business. And Based on the investigation the government agencies claimed Allys pricing and automobile dealer compensation system resulted in minorities (to whom I am an XXXX) paying higher interest rates than XXXX customers. Ally did not admit to these allegations but agreed to enter a settlement with the government

agencies in XXXX XXXX. In XXXX XXXX and XXXX XXXX Ally financial reported my account 30 days late and 60 days late to the XXXX major credit bureaus XXXX, XXXX, and XXXX. This not only hurt my credit but also financially as ally was allegedly charging me a higher interest than XXXX. So when I finally refinanced the above vehicle loan with my local credit union XXXX XXXX. I owed more on the vehicle than if Ally would have charged me at a lower interest.”

Is_discrimination: Yes

Confidence_score: 0.9

Reason: The complaint narrative indicates potential discrimination related to interest rate practices by Ally Financial. The complainant references a government investigation that found Ally’s pricing system resulted in minorities paying higher interest rates, and the complainant identifies as a minority. The complainant also describes financial harm due to allegedly being charged a higher interest rate than others, which aligns with interest rate discrimination. The narrative suggests a comparison between the complainant’s treatment and that of non-minority customers, indicating perceived unfair treatment based on race or ethnicity.

Complaint #1324587

Narrative: "I was contacted by Halsted financial services. I asked if they could send me a copy of the paperwork that determined I hold the debt. They refused to provide me with the paperwork as well as hung up on me. I called back XXXX 4 times after being hung up continuously. After finally speaking with someone and requesting the documents again I was told no multiple times and that I would experience wage garnishments as well as being sued in court if I did not settle the debt immediately. I again asked for the proper documentation that showed I owed the debt the representative refused to send me the paperwork. I asked to speak with their supervisor and was immediately told no that is not possible at this time. I said I will have nothing further to discuss with you until speaking with your supervisor I was hung up on. I’ve been called back and got the company name to file the complaint."

Is_discrimination: No

Confidence_score: 0.8

Reason: The complaint describes issues with debt collection practices, including refusal to provide documentation and threats of wage garnishment. However, there is no mention

or implication of discrimination or unfair treatment based on protected characteristics or comparisons to others. The complaint focuses on procedural issues and aggressive collection tactics rather than discriminatory practices.

The first case represents clear racial discrimination in lending, where the minority borrower was systematically charged higher interest rates compared to non-minority customers, supported by regulatory investigation findings. In contrast, the second case, while showing poor customer service and aggressive collection practices, does not indicate any differential treatment based on protected characteristics.

Appendix E The Impact on Asian Hate During Covid Period

The rise of Asian Hate, which began in early 2020 and persisted through 2021, originates from multiple factors including perceptions about COVID-19’s source, widespread xenophobia, and increasing economic pressures. This section investigates how disclosure policies affect Asian consumers during the COVID-19 pandemic. Throughout this period, Asians experience persistent discrimination and acts of violence (Jeung et al. (2021)), which directly affect the performance of Asian professionals in finance. For instance, Agarwal et al. (2023) finds that mutual funds managed by Asian-descended fund managers underperform relative to those managed by non-Asian counterparts. Similarly, Qiao et al. (2023) demonstrates that East Asian analysts produce financial forecasts with reduced boldness, greater pessimism, lower frequency of updates, and decreased timeliness during the pandemic. At the consumer level, individuals of Asian heritage may also face unfair treatment in financial markets. The CFPB’s complaint disclosure system potentially alleviates some effects of Asian Hate on consumers by exposing instances of discrimination. Building on the baseline framework established in this paper, we examine deposit rates using the following equation:

$$Y_{i,c,t} = \alpha + \beta_1 Treat_i \times Post_t \times Asian_c \times Covid_t \\ + X\theta + \mu_{Branch} + \mu_{YearQuarter} + \mu_{Institution \times Year} + \epsilon_{i,l,t}$$

Where Y_{ict} represents the deposit rates of branch i in county c at time t . $Treat_i$ is a binary variable that equals one if branch i belongs to an institution under CFPB oversight in the first quarter of 2015. A zero value of $Treat_i$ indicates that branch i is not subject to CFPB regulation. $Post_t$ captures the period before and after the CFPB made complaint narratives public, taking a value of one if the time is in the second quarter of 2015 or later. We classify counties where branches are located into two groups based on the Asian population ratio. If this ratio exceeds the median, the dummy variable $Asian_c$ equals one; otherwise, it equals zero. $Covid_t$ identifies the period before and after the pandemic, equaling one if the time is in or after January 2020. \mathbf{X} is a vector containing the remaining interaction terms from our setting and county characteristics, including population and per capita income growth rates. We also control branch, year-quarter, and institution-year fixed effects.

Most results in Table A10 indicate that the positive effect of complaint narratives disclosure on the Asian gaps in deposit rates at CFPB-regulated institutions strengthened during

the COVID-19 period. For example, Column (1), which examines the widely used deposit product, 12MCD10K, demonstrates an increase of 7.8 basis points in this effect. These institutions, considering factors such as reputation maintenance in this period, offered more favorable deposit rates to Asian consumers.

Table A10. Impact of CFPB Disclosure on Racial Gaps for Asians During the Covid Period

This table presents the results on the impact of CFPB disclosure on racial gaps for asians during the covid period in deposit interest rates. Columns (1) to (4) present the results for 12MCD10K, 12MCD10K and MM25K, CDs (four mainstream CD products), CDs and MMs (four mainstream CD products and four mainstream MM products), respectively. All columns control for demographic variables at the county level, including the logarithm of the total population and the income per capita growth rate, using the annual population of each county for weighting. We also control for branch, year-quarter, and institution-year fixed effects. Additionally, in Columns (2) to (4) we control for product fixed effects. Throughout the estimation process, we include all interaction terms involved in this setting. The key estimated coefficient, $Treat \times Post \times Asian \times Covid$, denotes the impact of disclosure on deposit rates in Asian communities. Standard errors, shown in parentheses, are clustered at the institution, county, and year level. The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively.

	(1)	(2)	(3)	(4)
	Deposit Rate			
VARIABLES	12MCD10K	12MCD10K and MM25K	CDs	CDs and MMs
$Treat \times Post \times Asian \times Covid$	0.078*** (0.023)	0.029*** (0.007)	0.061*** (0.016)	0.018 (0.013)
Branch FE	YES	YES	YES	YES
Product FE	-	YES	YES	YES
Year-Quarter FE	YES	YES	YES	YES
Institution-Year FE	YES	YES	YES	YES
Observations	306,974	569,897	1,175,974	2,119,692
R-squared	0.879	0.714	0.848	0.735

Appendix F Details for Online Ranking of Racial Gaps in Financial Services

Our research demonstrates that disclosure policies play a vital role in enhancing minority welfare in financial services by fostering transparency and accountability. In alignment with [Li \(2023b\)](#), we have expanded the platform with a ranking system that highlights how financial institutions serve different communities, particularly emphasizing the disparities between minority and white areas. This platform, building on a series of prior studies, already includes a comprehensive ranking of mortgage lenders. Our latest extension significantly broadens its scope, offering deeper insights into how branches cater to diverse communities while advancing transparency and equity in financial services.

Specifically, we have introduced new functionalities that rank financial branches based on disparities in four key financial outcomes: deposit rates, maintenance fees, auto loan rates, and credit card rates. These rankings are generated by comparing branches located in minority areas to those in white areas, controlling for branch, year-quarter, state, and other fixed effects to accurately capture the impact of racial composition in each location. The core of our analysis lies in identifying whether financial institutions treat consumers differently based on the racial composition of the communities they serve. By analyzing disparities in these four financial outcomes, we assess if branches located in minority communities offer systematically different financial products compared to those in white communities. Each regression model includes data at the branch level and applies fixed effects to control for time-specific economic conditions, product variations, and location-level regulatory differences, ensuring robust estimation of β .

$$\begin{aligned} Outcome_{i,l,t} = & \alpha + \beta Minority_l \times InstitutionID_i + X\theta \\ & + \mu_{Branch} + \mu_{YearQuarter} \\ & + \mu_{State \times Year} + \mu_{Product \times Year} + \epsilon_{i,l,t} \end{aligned}$$

In this equation, i , l , and t represent the branch, location (ZIP code), and year-quarter, respectively. The *Outcome* variable captures one of four financial indicators: deposit rate, maintenance fee, auto loan rate, or credit card loan rate, offered by branch i in location l during year-quarter t . The key independent variable, *Minority*, equals one if the branch op-

erates in an area where the minority population exceeds the median value based on the 2010 census. *InstitutionID* represents the financial institution to which branch i belongs. The interaction term, $Minority \times InstitutionID$, enables us to analyze branch-level variations across different communities within the same financial institution. The coefficient β reflects the racial disparity in outcomes between branches serving minority and white communities. A positive β suggests that branches in minority areas may charge higher auto loan and credit card loan rates, as well as higher maintenance fees, while a negative β would indicate lower deposit rates in these areas, suggesting potentially less favorable savings options.

Figure [A6](#) illustrates a country-wide ranking of branches based on deposit rates. These rankings aim to provide greater transparency into how branches price their deposit and loan products in different communities. By offering an accessible interface for the public, we hope that this initiative will contribute to increasing awareness of financial disparities and, in the long run, support efforts to create a more equitable financial environment.

Figure A6. Example of Online Ranking of Racial Gaps in Financial Services

This figure presents a screenshot extracted from the website, ranking 492 financial institutions across the country based on racial disparities in deposit interest rates. The table consists of three columns. Column (1) shows the Coefficient, which represents the difference in interest rates received by depositors in minority communities compared to those in white communities at each institution, with more negative values indicating greater disparities. Column (2) displays the Significance, which indicates the statistical significance of the coefficients (The symbols *, **, and *** correspond to significance at the 10%, 5%, and 1% levels, respectively). Column (3) identifies the financial institutions.

Compare the degree of discrimination
in deposit rates (Country-Wide)

Here are 492 options for you Ascending

We arrange the results in ascending order based on the size of the coefficients, by default
The numerical value of the coefficient indicates how much less interest minority depositors receive compared with white depositors at this bank.

Coefficient	Significance ⓘ	
-0.817	***	Jax Federal Credit Union
Coefficient	Significance ⓘ	
-0.709	***	American Heritage Federal Credit Union
Coefficient	Significance ⓘ	
-0.684	***	Spokane Teachers Credit Union
Coefficient	Significance ⓘ	
-0.631	***	Frost Bank

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