

Philadelphia Then and Now: Disparities in Mortgage Lending



Mortgages are the financial cornerstone that allows most American families to purchase their homes. However, some families face more difficulties than others when applying for a mortgage. Could applicants' race and ethnicity impact lending decision disparities?

In 2017, Reveal, the investigative journalism program, published findings showing that in the Philadelphia metropolitan statistical area (MSA), based on the Home Mortgage Disclosure Act (HMDA) data for 2016, Black and Hispanic applicants were 2.7 and 1.8 times, respectively, more likely to be denied conventional loans than Non-Hispanic White applicants. These results controlled for credit and census tract variables available in the HMDA filings. However, critical credit variables not available in HMDA data such as FICO credit score or other internally developed ones were not included in their analysis. This paper assesses the impact of missing confounders like FICO in the Reveal analysis for the 2016 HMDA data and provides the trend of the disparities from 2017 to 2022. Performing sensitivity analysis to determine if a missing confounders can explain away the race/ethnicity effect is a novel approach that has never been addressed within Fair Lending academic literature.



Reveal's story focused on the uneven access to the conventional mortgage market for applicants of color compared with White applicants. In particular, Reveal focused on Philadelphia because the race/ethnicity consistently proved to be statistically significant regardless of which variables were included in the regression model. Reveal obtained publicly available HMDA data. In addition, Reveal looked at every variable in the HMDA data set and added additional ones based on the feedback from experts and research. However, critical credit variables such as FICO scores were not available and were not included in the Reveal study. Several studies stated that FICO scores are correlated with race. Omitting such an important unmeasured confounder may bias the estimation of the true causal effect; thus, leading to wrong conclusions. Therefore, it is reasonable to ask if the results of Reveal are biased due to this omission: Is an unmeasured confounder such as FICO score strong enough to explain away the estimated race effect? The response to this question involves sensitivity analysis to unmeasured confounders. VanderWeele and Ding provided a minimal threshold of the risk ratio between the exposure (e.g., being Black) and the outcome (e.g., being denied a loan) for an unmeasured confounder (e.g., FICO score) to explain away the exposure effect on the output. This threshold is termed E-value, which is defined as the minimum strength of association on the risk ratio scale that an unmeasured confounder would need to have with both the treatment and the outcome to fully explain away a specific treatment-outcome association, conditional on the measured covariates.

In this paper, a logistic regression was fitted with the 2016 HMDA data. The E-value was derived to express the exposure-confounder (RR EU) and the confounder-outcome association (RR UD) needed to explain away the race effect on the denial probability. The same analysis was performed from 2017 to 2022.

In addition, this paper includes a popular disparity metric, the marginal effect, which provides the difference between the average probability of denials between minorities and White applicants after adjusting for the measured covariates.

Figure 1 and Figure 2 provide the trend of the rate of denials for conventional loans and refinancing by race. Both refinancing and conventional loans denial rates have been trending down since 2017 for all minorities and non-minorities.

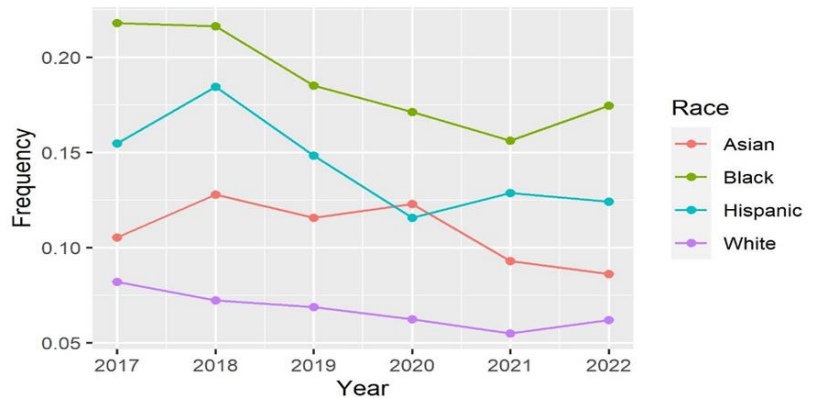


Figure 1: Percent of denied conventional loans by race and year

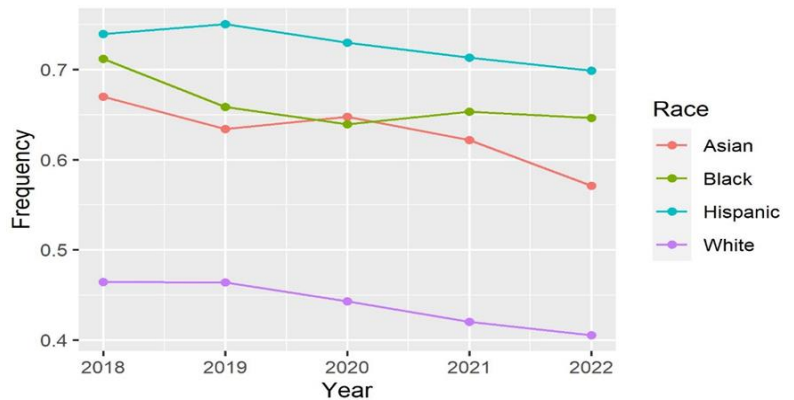


Figure 2: Percent of denied refinancing by race and year

As described above, Reveal fitted a logistic regression to the Philadelphia metro using HMDA data where the output or dependent variable is the binary decision, the exposure variables are minorities race, and the reference race is White. The results show that the odds of being denied a conventional loan in the Philadelphia MSA for Black applicants were 2.77-times higher than those of the White applicants. They were 1.78 times higher for Hispanic applicants and 1.41 times higher for Asian applicants.





Reveal analysis did not include applicants credit scores because they are not available in the HMDA data. Generally, White, and Asian applicants have, on average, higher FICO credit scores. Omitting such an important confounder may bias the results of the Reveal’s regression and may lead to wrong conclusions.

The E-value is a measure related to the “evidence for causality” in observational studies when they are potentially subject to confounding. The E-value is defined as the minimum strength of association on the risk ratio scale, which unmeasured confounders would need to have with both the treatment and the outcome to fully explain away a specific treatment-outcome association, conditional on the measured covariates.

As an example, in this paper, we assumed that the relative risk of FICO score below 683 (where denials are more likely to occur) on the outcome was 9 , which means applicants with FICO score below 683 are 9 times more likely to be denied the loan than those with FICO score above 683. In this scenario, we derive the relative risk of FICO on the exposure, for each minority (RR_{EU}) necessary to explain away the effect of minority race on the denial decision.

Minority	RR_{ED}	E-Value	RR_{UD}	RR_{EU}
Black	2.8	5.05	9	3.6
Hispanic	1.8	2.5	9	2
Asian	1.43	1.12	9	1.5

Table 1: Sensitivity measures of the Philadelphia metro disparity analysis for 2016.

The following figure shows the confounder strength frontier for different values of relative risk of the confounder on the outcome (RR UD) and the confounder on the exposure (RR EU). The intersection of the horizontal dashed line with the curve is the E-value. All the combinations of RR UD and RR EU that lie within the frontier and the purple region allow to explain away the race effect on denials.

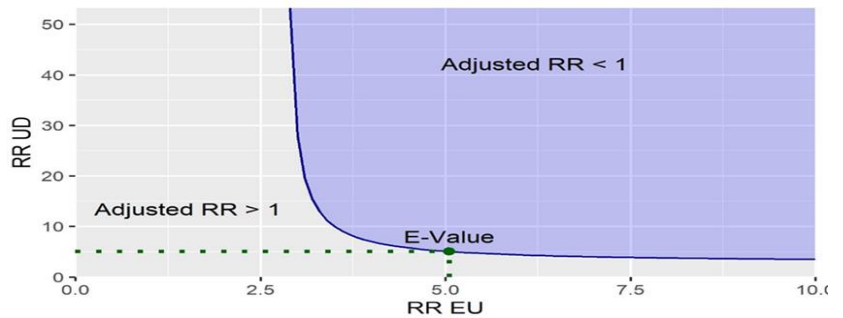


Figure 3: Confounder strength frontier

In addition, this study shows, from 2017 to 2022, the odds ratios, their 95% confidence bounds, the E-value, and the minimal value of RR EU when RR UD =9, that is required for the unmeasured confounder to neutralize the racial disparity of the conventional mortgage lending decision.

Variable	Year	Odds ratio	CI2.5	CI97.5	E-value	RR UD	Min. RR EU
Black	2017	2.742	2.248	3.339	4.927	9	3.505
Hispanic	2017	1.864	1.433	2.405	3.133	9	2.09
Asian	2017	1.331	1.082	1.627	1.994	9	1.388
Black	2018	2.243	1.821	2.76	3.914	9	2.656
Hispanic	2018	1.995	1.563	2.533	3.405	9	2.279
Asian	2018	1.921	1.583	3.324	3.252	9	2.171
Black	2019	2.467	1.987	3.058	4.368	9	3.02
Hispanic	2019	1.758	1.34	2.291	2.913	9	1.942
Asian	2019	2.053	1.687	2.492	3.524	9	2.365
Black	2020	1.512	2.266	3.425	5.02	9	3.589
Hispanic	2020	2.036	1.13	2.001	2.392	9	1.615
Asian	2020	2.668	1.657	2.493	3.488	9	2.339
Black	2021	1.765	2.202	3.23	4.777	9	3.37
Hispanic	2021	1.532	1.392	2.225	2.926	9	1.951
Asian	2021	1.532	1.255	1.863	2.435	9	1.641
Black	2022	2.591	2.089	3.211	4.621	9	3.234
Hispanic	2022	1.689	1.318	2.155	2.768	9	1.848
Asian	2022	1.44	1.193	1.823	2.317	9	1.571

Table 2: Odds Ratios and Sensitivity Measures of the Philadelphia Metro Disparity Analysis Over the Years

Regarding the marginal effects, the following figure shows the trends.

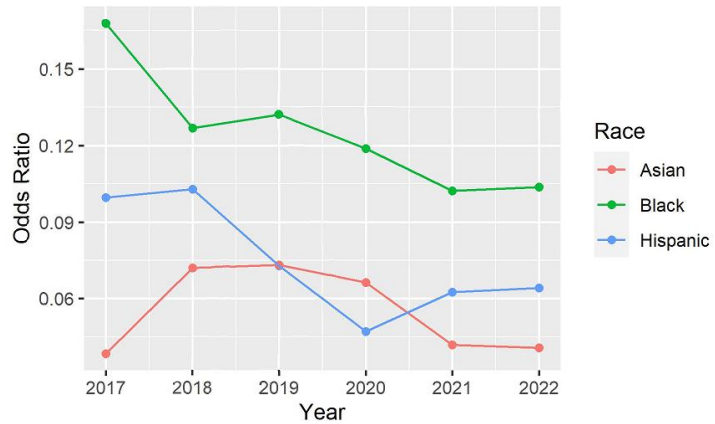


Figure 4: Trend of marginal effect for denial of conventional loans

As regards Black applicants, the marginal effect was 17% in 2017 and 11% in 2022. In other words, everything being equal, the probability of being denied a conventional loan in the Philadelphia metro for Black applicants was 17% higher than for White applicants in 2017 and 11% higher in 2022.

To summarize, in general, the lending disparities between minorities and similarly situated White applicants in the Philadelphia metro have been trending downward. However, for Black applicants, they are still at a significantly higher level and the sensitivity analysis seems to indicate that a very strong unmeasured confounder is needed in order to explain away this disparity.

In fact when the Reveal study came out, it triggered debates from the banks, some of them speculated that if FICO had been included in the model, the disparities would have been eliminated. Here, we show that may not have been the case.

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