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Racial climate and homeownership

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ABSTRACT

An important question aside from outright discrimination is whether poor underlying race relations in an area might create a chilling effect on homeownership for minorities. From 2012 onward, there were a series of high-profile events in the U.S. related to police brutality which highlighted racial tension. Using Google Trends, we characterize a locality's underlying racial climate based on search interest in these charged events. We use data from the American Community Survey prior to any of these flare-ups and show that the ownership decision for blacks is responsive to the racial climate; black homeownership in localities with the most charged racial climates is 5.6 percentage points lower than in the least charged racial climates based on a sample of movers.

1. Introduction

For more than 80 years, United States government policy has explicitly sought to promote homeownership. One way in which it has done so is by making homeownership a more lucrative investment. The tax code offers a multitude of advantages for homeownership, including the mortgage interest deduction, exclusion of significant portions of capital gains, property tax deductions, exclusion of imputed rental income, and implicit subsidization of interest rates through government-sponsored entities (Poterba and Sinai, 2008; Davis, 2012).

Why the preference for ownership? There are several conceptual arguments that relate to private or societal benefits. First, on the investment side, owner-occupied housing can be viewed as a hedge against rent risk.² In addition, homeownership has been found to increase wealth accumulation, often with magnitudes of approximately an additional \$10,000 in wealth per year of ownership (Turner and Luea, 2009). Second, some studies find that ownership is associated with more favorable outcomes for the family's children and the larger community. Haurin et al. (2002) find that ownership leads to a higher quality housing environment, greater cognitive ability and fewer child behavior problems.³ Other work examines positive spillovers from homeownership. DiPasquale and Glaeser (1999) find that homeowners invest more in social capital. There are also negatives associated with homeownership. Bayer et al. (2016) find that housing price risk is an important consideration and that minorities who purchased during the housing boom were especially vulnerable to economic fluctuations. In addition, Bostic and Lee (2008) highlight the risks and costs of "failed homeownership" among low- and moderate-income borrowers.

The benefits—whether causal or not—have created policy interest in the racial gap in homeownership. In 2002, President George W. Bush said, "We must begin to close this homeownership gap by dismantling the barriers that prevent minorities from owning a piece of the American dream."⁴ Shapiro (2006) argues for homeownership as a main strategy for closing the overall racial wealth gap. Despite these calls, Census Bureau data shows persistent gaps in ownership between whites and blacks of 25 percentage points for the past two decades. Despite major swings in the economy, white ownership has never fallen below 67% while black ownership has never exceeded 50%.⁵

One cause of this ownership gap is outright, illegal discrimination in housing and mortgage lending markets. A voluminous literature explores these issues. This type of discrimination represents a restriction in the supply of housing for blacks. In one recent audit study using paired test subjects, black homebuyers were informed about and shown roughly 17% fewer homes than white homebuyers (Turner et al., 2013). There are also concerns about geographic steering and discrimination known as "redlining" (Tootell, 1996; Ondrich et al., 2001; 2003; Ross and Tootell, 2004). Evidence on lending discrimination reveals

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Received 31 May 2017; Received in revised form 12 December 2017; Accepted 13 December 2017 Available online xxx 1051-1377/ © 2017. that minorities are more than twice as likely to be denied a mortgage as whites, although correcting for omitted variables bias significantly diminishes the impact of race (Munnell et al., 1996).

Our work focuses on the impact of a locality's overall "racial climate" on the decision of blacks to own homes. Racial climate would include both factors that affect the supply of housing to blacks such as housing and mortgage discrimination, but additionally demand-side factors that influence the decision to "plant one's roots" and invest in a community. Obvious factors would include labor market discrimination, unequal educational opportunity, racism, and policing.⁶ We view "racial climate" in our setting as parallel to "chilling effects" in other recent work. For example, in the context of the 1996 U.S. welfare reform which included anti-immigrant language, the general policy environment can matter for decision making apart from the formal rules, and such indirect effects are termed chilling effects (Watson, 2014).⁷ Such chilling effects are inherently difficult to measure, and researchers attempt to find proxies for the overall climate.⁸

In our context, since virtually all standard microdata is wholly inadequate for measuring racism or racial climate spatially (and likely subject to misreporting), we follow an approach pioneered by Stephens-Davidowitz (2014) in using Google Trends. In this study, racial animus at the state-level was proxied by searches related to racial epithets. In our approach, we use a variety of search terms and topics related to "Police Brutality" to measure the long-run state of race relations by locality. In particular, the 'Black Lives Matter' movement was formed in the aftermath of the shooting of 17-year-old Trayvon Martin by a private citizen in February 2012.9 Other high profile incidents involving blacks and the police (rather than private parties) include the shooting of 18-year-old Michael Brown in Ferguson, Missouri in 2014, the shooting of 12-year-old Tamir Rice in Cleveland, Ohio in 2014, and the death of 25-year-old Freddie Gray in Baltimore, Maryland in 2015. Our work uses Google search interest related to these high-profile policing events occurring in 2012 onward as a proxy for a locality's racial climate, where heightened interest in such topics is arguably associated with a more charged racial climate. Drawing upon data from the American Community Survey (ACS) prior to these events occurring, we examine the ownership decision among a large sample of recent movers. After controlling for other factors, we find that homeownership of blacks in the most racially charged localities is 5.6 percentage points lower than in the least charged localities.

2. Data description

We use the ACS, a nationwide survey administered by the Census Bureau that asks detailed questions about population and housing characteristics, as our principal data source. The ACS samples approximately one percent of the U.S. population; we use respondents in the years 2005 to 2011, prior to the high profile incidents used to measure race relations. Like the Decennial Census, participation in the ACS is mandatory, and the survey can be completed online or by mailing in a paper questionnaire. The ACS identifies all 50 states and the District of Columbia and additionally identifies Public Use Microdata Areas (PUMAs)—approximately 2300 areas of at least 100,000 people nested entirely within a state. The ACS contains sufficient information to identify localities, which we map into metro areas in a similar fashion as in Courtemanche et al. (2017).

The primary variable of interest, Racial Climate, is derived from Google Trends data. Google data, which aggregates millions of searches, provide insights into social perceptions that are hard to accurately elicit from survey data (Stephens-Davidowitz, 2017). Surveys, such as the widely used General Social Survey, which seek to understand concerns and attitudes are wholly inadequate at analyzing racial climate at a metro area level due to insufficient sample sizes, lack of fine geographic locations, and concerns about reporting. Researchers have used Google data in a wide variety of contexts such as studying the influence of racial animus on elections (Stephens-Davidowitz, 2014), the incidence of child abuse during the Great Recession (Stephens-Davidowitz, 2013), and the user base of Bitcoin (Yelowitz and Wilson, 2015).

Google data is available at the Designated Market Area (DMA) level, which we map into metro areas.¹⁰ We focus on term/topics related to police brutality. Interdisciplinary studies, such as Chaney and Robertson (2013), take the view that such policing events reflect racism and discrimination, as well as greater range of social problems including racial profiling and harsh treatment in the criminal justice system. In addition, Fryer (2016) finds that police use of force is greater for blacks relative to whites. Racial climate based on search interest in police brutality represents racial tension at an institutional level, which arguably captures race relations better than the use of racial slurs like in Stephens-Davidowitz (2014). Furthermore, racially charged areas could also be associated with heightened discrimination, which further influences housing decisions. To gauge the racial climate we create an average Z-score index using the following search terms/topics: Police Brutality, Black Lives Matter, Shooting of Michael Brown, Ferguson Unrest, Trayvon Martin, Death of Freddie Gray, and Shooting of Tamir Rice.¹¹ Fig. 1 illustrates the average Z-scores for the racial climate in the metro areas used in the analysis.

Appendix Table A.1 further illustrates the variation in the aggregate index by metro area. Several of the metro areas with the largest index for racial climate are from areas where the incidents occurred. Arguably, the incidents occurred in these metro areas because of heightened racial tensions and the flare-ups would not have necessarily occurred in different areas for the same stimulus. By aggregating search interest across several different events—that should not be correlated otherwise—we mitigate the influence of each individual event. Nonetheless, there is concern that these areas received a disproportion-

² Sinai and Souleles (2005) find that the probability of homeownership increases faster with rent volatility for long-horizon households than for short-horizon households.

³ However, Holupka and Newman (2012) argue that such beneficial homeownership effects may be due to selection bias.

⁴ See http://www.presidency.ucsb.edu/ws/?pid=25063.

⁵ See https://www.census.gov/housing/hvs/files/annual16/ann16t_22.xlsx.

⁶ See https://www.nytimes.com/2016/08/21/us/ milwaukee-segregation-wealthy-black-families.html for an example of black households locating in less racially charged areas leading to segregation. Presumably, racial climate also influences the decision to locate/invest in homeownership across cities and not just within cities.

⁷ Other examples include internet use by Muslim-Americans in the aftermath of the September 11th attacks (Sidhu, 2007) and college applications following affirmative action bans (Antonovics and Sander, 2013).

⁸ Watson (2014) proxies for the chilling effect on Medicaid participation for children of immigrants using spatial and temporal variation in federal enforcement actions from the Immigration and Naturalization Service.

⁹ See http://blacklivesmatter.com/herstory/

¹⁰ There are a total of 210 DMA in the U.S., which correspond to different media markets as defined by Nielsen. We use Sood (2016) to assign DMA information to counties and use a crosswalk from Missouri Census Data Center (2012) to translate county level information into metro areas. Appendix Table A.5 lists the final crosswalk between DMA and CBSA. In addition, a downloadable version of our constructed crosswalk can be accessed at http://www.yelowitz.com/racialclimate. For our sample, an average of roughly two metro areas map into a single DMA. In the analysis, we cluster at the DMA level.

¹¹ We use the average indexes for these measures over time rather than exploiting any time variation in the metrics. The average Z-score is created by subtracting the mean and dividing by the standard deviation for each of the search terms. We then sum the scores and divide by the standard deviation of the sum to get an index of mean zero and standard deviation one (Chetty et al., 2011; Kling et al., 2007).

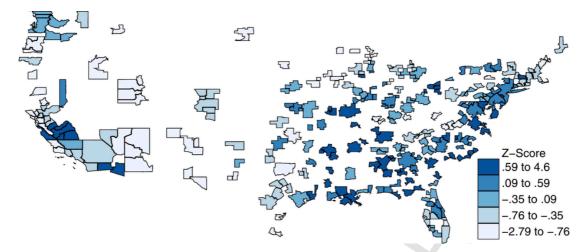


Fig. 1. Google Trends: Racial Climate Z-score by Metro Area. Note: Only CBSAs used in the analysis are shown, which represent 82.7% of the population in 2010. Z-scores are translated from DMA information provided from Google Trends.

ate amount of search interest because the incident occurred in that locality. Consequently, in a robustness check, we exclude St. Louis, Missouri; Baltimore, Maryland; Cleveland, Ohio; and Orlando, Florida from the analysis.

The regression analysis will evaluate the relationship between racial climate and black homeownership. Anecdotally, Salem, Oregon, and Jacksonville, North Carolina provide an example of a negative relationship between racial climate and homeownership. Oregon with one of the best racial climates has a black homeownership rate of 39.7% while Jacksonville, North Carolina has one of the worst racial climates has a black homeownership rate of 32.7

The earliest incident that we use to gauge underlying racial climate was the "Shooting of Trayvon Martin" that occurred on February 26, 2012.¹² To disallow the event itself from driving homeownership rates (i.e., decreased black homeownership due to actual violence/destruction), we focus on a sample from the ACS that entirely predates any of the actual events, using the years 2005–2011. The racial climate metric is based on the assumption that search interest in these events is a manifestation of latent racial tension in an area rather than the event itself driving homeownership rates.

In addition to Google Trends data, we use other metro level data to account for factors that might influence the housing decision. These data include monthly Fair Market Rents (FMR) from the Department of Housing and Urban Development,¹³ Housing Price Index (HPI) from the Federal Housing Finance Agency (FHFA),¹⁴ and data from the FBI Uniform Crime Report statistics.¹⁵ Furthermore, we calculate the income to poverty ratio, share in manufacturing, and percent black using the ACS and the logarithm of population from the 2010 Decennial Census for each metro area.

We restrict our sample to (a) households whose head is either black or white (non-Hispanic) (b) heads who reside in metropolitan areas¹⁶ (c) metro areas that have information on monthly FMR from the Department of Housing and Urban Development, HPI from the Federal Housing Finance Agency, and crime statistics as reported in the FBI Uniform Crime Report. The 329 metro areas used in the analysis contain 82.7% of the U.S. population as reported in the 2010 census.

To gauge the influence of racial climate on black homeownership, we analyze three different samples of households. First, we analyze the full sample of households who reside in a metro area. However, use of the full sample raises concerns with timing of measurement and reverse causality. For our principal outcome—homeownership—it is important to recognize that the vast majority of households are established in a location and plausibly made their homeownership decision at a time in the past that reflects a different racial climate than the present climate.¹⁷ For example, ACS tabulations indicate that nearly 56% of all homeowners in 2011 lived in their residence for 10 or more years. Therefore, the measure of racial climate derived from Google Trends data likely does not capture the characteristics of the environment that drove the rent versus own decision for a significant majority of the full sample. In addition, the homeownership rate in a metro area could influence search interest and the racial tension variable resulting in reverse causality.¹⁸

As one alternative, we analyze a younger sample (household heads aged 18–35) who when surveyed likely made the homeownership decision in the recent past. For this sample, the racial climate captured by the Google Trends index variable likely better characterizes the environment when the decision to rent or own was made. Furthermore, the homeownership of the smaller sample is less likely to drive the racial climate variable mitigating concerns of reverse causality.

As another alternative, we analyze households that moved in the year prior to being surveyed. This sample includes households that actively made the decision to rent or own in the racial environment captured by the racial climate index. We exclude households that moved within a metro area as they were exposed to the same racial climate in the year prior to their move and there is persistence in the homeowner-

¹² See Appendix Fig. A.1 for an illustration of the timing of search interest in each event/ topic.

¹³ We use data on 2-Bedroom Units at the county level. For counties with sub-areas reported, we weight the areas by their population in 2010 to aggregate to the county level. These are then mapped into metro areas. Data accessed from https://www.huduser.gov/portal/datasets/fmr.html.

¹⁴ For the HPI we use Metropolitan Statistical Areas and Divisions (Not Seasonally Adjusted) estimated using Sales Prices and Appraisal data. We average across quarters to get an annual measure and normalize the measure to be 100 in 2005. Data accessed from https://www.fhfa.gov/DataTools/Downloads/Pages/House-Price-Index-Datasets. aspx#mpo.

¹⁵ From the crime data, we create a crime Z-score calculated in a similar fashion as the racial climate index using statistics on violent crimes, murder/non-negligent manslaughter, robbery, aggravated assault, property crime burglary and motor vehicle thefts by metro area. See Fig. A.2 for a map of the crime Z-score.

 $^{^{16}\,}$ We exclude all micropolitan areas and areas and Public Use Micro Areas that do not map into a core-based statistical area (CBSA).

¹⁷ Economic theory predicts that households should only respond to changing racial climate in the short-run inasmuch the costs of a poor racial climate exceeds moving costs (Ihlanfeldt, 1981).

 $^{^{18}\,}$ For example, low rates of black homeownership could result in more searches related to the racially charged events.

ship decision. Using the sample of across-metro movers also mitigates concerns for reverse causality as the small proportion of households that moved are unlikely to influence Google searches enough to significantly impact the racial climate index.

Table 1 presents the basic summary statistics for the three samples described above. As shown, the racial composition of the three samples is relatively similar with blacks representing between 15% and 19% of the sample. In relation to the full sample, the across-metro movers sample is younger, less likely to be married, better educated, and significantly less likely to own a home (32% in relative to 68%). The young sample is also less likely to be married, less likely to have children, have lower income, and are significantly less likely to own a home in comparison to the full sample.

3. Empirical methodology

To test for the influence of racial climate on black homeownership, we estimate the following linear probability model, in the spirit of Watson's (2014) analysis of chilling effects of Immigration and Naturalization Service (INS) enforcement actions on non-citizens.¹⁹

$$Own_{ijt} = \beta_0 + \beta_1 Black_i \times Climate_j + \beta_2 Black_i + \beta_3 X_i + \beta_4 Local_{it} + \beta_5 Black_i \times Local_{it} + \delta_i + \delta_t + \epsilon_{iit}$$
⁽¹⁾

where Own_{ijt} is an indicator that household *i* owned a home in location j at time t rather than rent and $Black_i$ is an indicator that the head of the household is black. *Climate_i* is the time-invariant index for the racial climate that varies at the DMA level (higher values represent a worse racial climate). X_i measures characteristics of the head and other family members including age, gender, marital status, educational attainment, and number of children. Local_{it} measures factors that vary across cities and over time including FMR, HPI, Crime Rates, and percent in manufacturing.²⁰ Following Cutler and Glaeser (1997) we also include the interaction $Black_i \times Local_{it}$ to allow for differential location effects on blacks relative to whites. In addition, we include the interaction of race with other locality characteristics including logarithm of the population in 2010, income to poverty ratio, and percent black. δ_i and δ_t are fixed effects for locality and time. The specification does not include Climate, itself since it is subsumed with locality fixed effects. The locality fixed effects control for differences in levels for home prices, whereas the HPI controls for differences in growth of housing prices over time. Locality fixed effects also control for time invariant racial differences in the residential location inside a metro area, which influence homeownership rates (Deng et al., 2003). The error term ε_{ijt} is corrected for clustering at the DMA level.

Under the assumption that higher values of $Climate_j$ reflect a worse racial climate, we expect the coefficient β_1 —the interaction of a worse racial climate and black—to be negative. The coefficient captures at least two effects. First, black households may choose not to invest in a community with a poor racial climate and decide to rent instead. Second, households may select a location based on the racial climate. If this selection occurs, homeownership rates in communities with a good racial climate will be higher while simultaneously reducing the homeownership rate in communities with a poor racial climate.²¹ Therefore, β_1 can be interpreted as capturing the net effect of these two behaviors

Table 1 Summary statistics.

	Full sample	Young	AcrossMetro movers
Demographics Black _i	0.15	0.19	0.16
White _i	0.85	0.81	0.84
Male _i	0.53	0.49	0.54
Age 18–29 _i	0.11	0.55	0.26
Age 30–39 _i	0.17	0.45	0.23
Age 40–49 _i	0.21	0.00	0.20
Age 50–59 _i	0.21	0.00	0.15
Age 60–69 _i	0.14	0.00	0.09
Age $70 + i$	0.16	0.00	0.07
Family Structure Married _i	0.50	0.40	0.40
Widowed _i	0.10	0.00	0.06
Divorced _i	0.17	0.07	0.18
Separated _i	0.03	0.03	0.04
Never Married _i	0.20	0.50	0.33
Own Children: 0 _i	0.72	0.56	0.76
Own Children: 1 _i	0.12	0.18	0.11
Own Children: 2 _i	0.10	0.16	0.09
Own Children: $\geq 3_i$	0.05	0.10	0.04
Education and Income			
Less than HS Grad _i	0.10	0.06	0.05
High School Grad _i	0.26	0.22	0.16
Some College _i	0.31	0.37	0.34
Bachelor's Degree _i	0.20	0.26	0.28
Graduate Degree _i	0.12	0.10	0.16
Household Income _i (\$1k)	80.79	64.20	74.68
Housing and Area Owns	0.68	0.41	0.32
Home _i			
Monthly FMR_j (\$1)	978.92	955.58	996.44
FHFA HPIj	100.03	100.71	100.13
Crime Rates _j (Z-score)	0.34	0.35	0.44
Observations	5,885,509	1,017,563	150,478

Note: The sample includes white or black headed households from 370 CBSAs in years 2005–2011 in the ACS. Information is reported for household head *i*, and local (metro area) *j* using housing unit weights. Monetary values are reported in 2015 dollars. FHFA HPI is set equal to 100 in 2005.

(which work in the same direction). Identification comes from the assumption that the racial climate does not affect the investment/ownership decision of white households; therefore, our specification nets out other fixed local characteristics with δ_{j} . In addition to the selection of location, there is also selection in the decision to relocate. The estimate will not capture this effect, which could lead to an underestimation of the influence on racial climate on black homeownership.

4. Results

Table 2 presents the results from our analysis of the three main samples. The main variable is a control for residing in an area with a racial climate index in the highest quartile. We drop observations in the middle quartiles such that the comparison is between the top and bottom quartiles.²²

The first column shows no evidence that the racial climate negatively impacts either race's homeownership decision. Nonetheless, as outlined, the full specification analyzes the influence of the current racial climate on homeownership for individuals that made the decision under a presumably different climate and also is subject to concerns of reverse causality.

In the second column, we repeat the analysis for households headed by individuals aged 18–35 who likely made the homeownership decision under the racial climate captured by the racial climate index. The

¹⁹ Nearly identical results were obtained for the main specification using a Probit model rather than a linear probability model.

²⁰ Yelowitz (2007) and Yelowitz (2017) examine the impacts of house prices and rents at the local level over time using data from FHFA and HUD.

 $^{^{21}}$ The discussion of this second factor relies on the assumption that potential homeowners exhibit this behavior more than renters. Given the investment associated with homeownership, this is likely a reasonable assumption.

²² The main results are robust to conducting the analysis using the top and bottom terciles and quintiles. The coefficient on racial climate is larger in absolute magnitude (more negative) using the top and bottom quintiles and slightly smaller using the top and bottom terciles. The results are presented in Appendix Table A.2.

Table 2

Influence of racial climate on homeownership.

	Full	Young	Movers
$Black_i \times Aggregate$	- 0.004	042***	056***
Index GT Top 25%	(0.013)	(0.016)	(0.016)
Black	235***	273****	- 0.094
	(0.084)	(0.077)	(0.112)
Household Income _i	.079***	.210****	.126***
(\$100k)	(0.003)	(0.019)	(0.007)
Monthly FMR _{i, t} (\$1k)	.018***	0.010	072*
- ,,-	(0.006)	(0.027)	(0.042)
FHFA HPI _{i, t} (100)	.017**	0.036	- 0.010
<i></i>	(0.007)	(0.022)	(0.029)
Crime Rates _{i, t}	0.002	0.005	0.011
(Z-score)	(0.002)	(0.004)	(0.012)
Share	.093**	.201**	.588*
Manufacturing _{i, t}	(0.042)	(0.098)	(0.300)
$Black_i \times$			
Monthly FMR _{i, t}	.061***	.193***	.215***
(\$1k)	(0.021)	(0.057)	(0.061)
FHFA HPI _{i, t}	.090***	0.027	0.045
(100)	(0.027)	(0.025)	(0.046)
Crime Rates _{i, t}	.011***	.011**	- 0.002
(Z-score)	(0.004)	(0.005)	(0.011)
Percent Black _i	.186***	.287***	.367***
*	(0.054)	(0.059)	(0.126)
Income-to-Poverty	- 0.033	0.006	- 0.040
Ratio _i (100)	(0.023)	(0.033)	(0.053)
Log Population _j	0.143	- 0.706	- 0.970
(100)	(0.516)	(0.582)	(0.598)
Share	- 0.137	238*	- 0.239
Manufacturing _{i, t}	(0.097)	(0.125)	(0.232)
Obs.	2,089,624	367,195	53,115

Note: Dependent variable is homeownership. The full sample includes white or black headed households from the ACS in years 2005–2011. The sample is restricted to households that live in areas that fall into the upper and lower quartiles for the Aggregate Index of racial climate. Controls for age, education level, presence/number of children, gender/marital status of household head, and year fixed effects were included but not reported. Aggregate Index for Racial Climate uses the following Google Search Terms/Topics: Police Brutality, Black Lives Matter, Shooting of Michael Brown, Ferguson Unrest, Trayvon Martin, Death of Freddie Gray and Shooting of Tamir Rice. Standard errors are clustered at the DMA level and are shown in parentheses *** p < .01, ** p < .05, * p < .1.

results indicate that black households in the most racially charged metro areas are 4.2 percentage points (p-value < .001) less likely to purchase a home relative to those in the least racially charged areas from a base of 25.0% black homeownership.

The last column of Table 2 presents the results from the sample of across-metro movers. The results indicate that black households in the most racially charged metro areas are 5.6 percentage points less likely to purchase a home (from a base of 20.3% black homeownership).²³

Across the specifications, blacks are significantly less likely to be homeowners. Blacks are more likely to purchase a home relative to whites as FMR or the HPI increases. Higher crime rates are correlated with increased black homeownership while largely uncorrelated with white homeownership. The proportion of black households in a metro area is also associated with increased black homeownership. Although not presented in the table, the results are consistent with the standard findings that ownership rises with age, education level, marriage and presence of children.

A important consideration for the across-metro movers sample is the persistence in homeownership across moves due to preferences and

home equity. Although the ACS does not contain information on the previous homeownership status of movers, it does contain information on the household's original location. To account for differences in the likelihood that a household owned a home, we construct a measure of the average homeownership rate of non-movers in the location from which the household moved with similar characteristics. In constructing the measure, there is a trade-off of including many characteristics and having a precise match with small cell sizes or including a few characteristics but having large cell sizes. To determine which characteristics to use in the creation of the metric, we first estimate a simple model of homeownership on standard demographic controls.²⁴ To rank order the controls, we run the full model and obtain the R^2 and then run the model repeatedly leaving out individual variables (or groups of variables) one at a time noting the R^2 for each model. Based on the difference in explained variation in homeownership, we determine that the most influential variables are age and income. We construct the first measure using these two characteristics. We then create two more measures using progressively more of the demographics based on importance in explaining homeownership.²⁵ Fig. 3 shows the distribution of homeownership for non-movers based on demographic group and location using the second measure containing age, income, marital status, and race. As an example, the likelihood of homeownership in the previous location for household heads aged between 35 and 54, with income between \$30,000 and \$60,000, were married, and white was 31% in the Bronx of New York City and 78% in Nashville, Tennessee.

Table 3 analyzes the sensitivity of the results to controlling for the likelihood of homeownership prior to the move. The first column replicates the results from the across-metro movers presented in Table 2 but with the restricted sample originating from the exclusion of observations where there were insufficient data to estimate the likelihood of homeownership for a particular location and demographic cell. The last three columns sequentially add controls for the likelihood of homeownership keeping sample sizes constant for comparison. The estimates are extremely stable with the inclusion of additional controls indicating that our racial climate results are unlikely to be driven by omitted variable bias related to previous homeownership. Among across-metro movers, a charged racial climate reduces black homeownership by 9 percentage points controlling for the likelihood of previous homeownership.

5. Robustness

Table 4 contains the results from several different robustness checks using the across-metro movers sample with locality fixed effects and interactions of race and metro characteristics as the baseline model.²⁶ The first column excludes observations from the focal points of the events that were used in the derivation of our index of racial climate. In particular, we exclude the four "hot spots" of St. Louis, Missouri; Baltimore, Maryland; Cleveland, Ohio; and Orlando, Florida. This exclusion address concerns that search intensity in the area of the actual event might be elevated given the location rather than representing height-ened racial tensions. As shown, the magnitudes are not sensitive to the exclusion of these areas; even excluding these areas, black homeownership falls by 6 percentage points.

²³ These results are sensitive to the inclusion of metro controls by race interactions. In Appendix Table A.3 we present the results from a Gelbach Decomposition of the influence of each covariate on the racial climate index (Gelbach, 2016). The findings show that interaction of race with Percent Black is the primary covariate that cause the point estimate on the racial climate variable to change from a small insignificant negative to a statistically significant and more negative result when the interactions are added.

²⁴ The results are reported in Appendix Table A.4.

²⁵ Household observations are mapped into MIGPUMAS using Ruggles et al. (2015). We exclude observations where there are not at least 100 observations for the particular location/demographic cell. Overall, households are mapped into 26,497 unique cells based on demographics, and former locations. Homeownership rates are calculated using the ACS from 2005 to 2015.

 $^{^{26}\,}$ In addition, we include the first measure of the likelihood of previous homeownership (based on age and income) in each specification.

Table 3

Influence of previous homeownership.

	(1)	(2)	(3)	(4)
$Black_i \times$	_	_	_	_
	.093***	.086***	.086***	.086***
Aggregate Index GT	(0.027)	(0.029)	(0.029)	(0.029)
Top 25%;				
Likelihood Owned		$.232^{***}$	$.233^{***}$.233***
Home,				
Former location _{p, a, I}		(0.033)	(0.059)	(0.058)
Likelihood Owned			_	_
Home,			0.001	0.047
Former location _{p, a, I,}			(0.086)	(0.135)
m, r				
Likelihood Owned				0.047
Home,				
Former location _{p, a, I,}				(0.097)
m, r, k, e				

Note: Results are presented for analysis of household head *i*, cbsa *j*, Migration PUMA *p*, age group *a*, income bin *I*, marital status *m*, race *r*, children status *k* and education *e*. The sample is restricted to households that moved to a new metro area that fall into the upper and lower quartiles for the Aggregate Index of racial climate. CBSA fixed effects and all variables used in Table 2 were included but not reported. Standard errors are clustered at the DMA level and are shown in parentheses ^{***} p < .01, ^{**} p < .05, * p < .1.

A possible concern with the metric used to measure racial climate is that it might be correlated with other forms of social activism. In the second column, we add an interaction between the Z-score for Google searches of "Climate Change" and $Black_i$ to control for possible correlations between social activism and search interest in incidents related to racial climate. The results are nearly identical to the main specification that excludes the interaction for climate change. This indicates that our metric of racial climate is unlikely to be driven by social activism rather than increased racial tensions.

In the third column, we present results from a specification that account for the racial climate in the former location. Previous racial climate presumably influenced the homeownership decision. Through persistence in homeownership, the former racial climate could impact homeownership in the new location and consequently bias the results. The point estimate for the specification increases in absolute magnitude but the difference is driven by the sample and not the inclusion of the added control.²⁷ Furthermore, as in the full sample of households, the timing of the measurement for the previous racial climate's homeownership decision likely will not reflect the racial climate when they made the decision to rent or buy a home in the previous location.

Another potential concern in that the racial climate index could be picking up the influence of other features of an urban area such as racial segregation that differentially impact black households. To control for differences in racial composition/segregation within metro areas, we construct a dissimilarities index following Cutler and Glaeser (1997). We use 2010 Decennial Census data with racial counts at the census tract level, which proxies for neighborhoods.²⁸ Housing segregation in a metro area is defined as:

Housing Segregation =
$$\frac{1}{2}\sum_{i=1}^{N} \left| \frac{Black_i}{Black} - \frac{White_i}{White} \right|$$
 (2)

where Black_i (White_i) is the number of blacks (whites) in census

tract *i* and *Black* (*White*) is the number of blacks (whites) in the metro area. If blacks are evening distributed across the metro area, then the index would be zero. If there is complete segregation, then the index would be one. As reported in Table 4, the inclusion of this measure interacted with race does not alter the main finding and the interacted measure itself is not statistically significant. This finding indicates that racial segregation does not differentially impact blacks' rent versus own decision relative to whites.

As we are using search interest in very contemporaneous events the racial climate might be different for the earliest years of our sample. To test for this possibility, we split the sample between the earlier years (2005–2008) and the years directly preceding the Black Lives Matter movements (2009–2011). The last two columns of Table 4 present the results and once again there is a consistent negative relationship between the likelihood of homeownership for blacks and a poor racial climate. The results, however, are weakly statistically significant in the earlier sample but strongly statistically significant in the later sample. These findings provide greater support for looking at households in transition (movers) rather than the full sample as the racial climate or the impact of racial climate appears to change over time.

Lastly, we analyze the importance of the events included in the construction of the racial climate index. We run the main specification of movers repeatedly excluding one of the search topics from the construction of the index (leave-one-out approach) to analyze the influence on the main result. Table 5 presents the findings. There is still a significant impact of the racial climate on black homeownership in five out of six specifications, however, for the exclusion of the events surrounding Ferguson Missouri the coefficient becomes statistically insignificant. These results highlight some variability in the magnitude of the effect of racial climate depending on the inputs into the racial climate variable.

6. Conclusion

Innovations in creating data and measuring sentiment—via Google Trends—has opened up new possibilities for examining important issues, such as the role for chilling effects on behavior. The costly decision to own—and subsequently invest more in a community—is likely related to the community's amenities and disamenities. We show that negative race relations—as represented by public interest in well publicized policing incidents—significantly reduces minority home ownership in a community. The results vary by specification, but our preferred specification shows that blacks in the most charged racial climate purchase homes 5.6 percentage points less than those who reside in localities with the least charged racial climate from a base of 20.3% black homeownership. Not only does this imply that these households are not receiving the benefits of homeownership, but it also implies that black households are less likely to invest in their communities. (Fig. 2)

Our results, insofar as they capture problems with the criminal justice system, suggest that some recent proposals with bipartisan support to reform policing and sentencing may have larger social benefits beyond those directly aggrieved. Reforms in police tactics—such as additional training, body cameras, and the use of outside agencies to investigate misconduct—have broad-based support (Ekins, 2016). The results suggest that efforts to reform police conduct could have the positive spillover of greater community investment. Furthermore, inasmuch as homeownership increases wealth accumulation, these policy reforms could help mitigate the overall racial wealth gap.

Appendix A

 $^{^{27}\,}$ The results from the specification without the control for the interaction of race and the previous racial climate restricted to the same sample yields a statistically significant point estimate of - 8.2 percentage points.

²⁸ We use Summary File 1 (SF1) Urban-Rural Update files accessed through http://www.ciser.cornell.edu/pub/2010SF1/census2010sf1.shtm.

Table 4

Influence of racial climate on homeownership, robustness checks.

$Black_i \times$	Hot Spots Excluded 060**	Climate Change – .054***	Former racial climate – .079 ^{***}	Segregation Index 047**	2005– 2008 – .049*	2009– 2011 – .045 ^{***}
Aggregate Index GT Top $25\%_j$ Black _i ×	(0.024)	(0.018) 0.007	(0.022)	(0.019)	(0.028)	(0.015)
Climate Change GT _i		(0.011)				
$Black_i \times$			-0.001			
Aggregate Index GT _p			(0.004)			
$Black_i \times$				- 0.055		
Segregation Index _i				(0.083)		
Obs.	48,026	53,003	33,149	52,287	35,301	17,702

Note: The sample is restricted to households that moved to a new metro area that fall into the upper and lower quartiles for the Aggregate Index of racial climate. Furthermore, the sample is restricted to households that moved to a new metro area. CBSA fixed effects and all variables used in Table 2 and the first measure of the likelihood of previous homeownership were included but not reported for household head *i*, cbsa *j*, and Migration PUMA *p*. Standard errors are clustered at the DMA level and are shown in parentheses *** p < .01, ** p < .05, * p < .1.

Table 5

Sensitivity to events used in index, leave-one-out.

Excluded event:	Police	Black Lives	Ferguson/	Trayvon	Freddie	Tamir
	Brutality	Matter	M. Brown	Martin	Gray	Rice
$Black_i \times$						
Aggregate Index1 GT	055****					
Top 25%;	(0.015)					
Aggregate Index2 GT		032*				
Top 25%;		(0.018)				
Aggregate Index3 GT			- 0.029			
Top 25%;			(0.021)			
Aggregate Index4 GT				042***		
Top 25% _j				(0.016)		
Aggregate Index5 GT					072^{***}	
Top 25% _j					(0.014)	
Aggregate Index6 GT						050***
Top 25% _j						(0.013)
Obs.	60,962	56,778	60,035	68,691	56,190	59,267

Note: The sample is restricted to households that moved to a new metro area that fall into the upper and lower quartiles for the Aggregate Index of racial climate. Furthermore, the sample is restricted to households that moved to a new metro area. CBSA fixed effects and all variables used in Table 2 and the first measure of the likelihood of previous homeownership were included but not reported for household head *i*, and cbsa *j*. Standard errors are clustered at the DMA level and are shown in parentheses ^{***} p < .01, ^{**} p < .05, * p < .1.

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Table A.1

Google trends, race relations indexes (sorted desending based on all indices).

	Avg	Police	Black lives	Michael	Ferguson	Trayvon	Freddie	Tam
Metro area	index	brutality	matter	Brown	unrest	Martin	Gray	Rice
St. Louis, MO-IL	49	24	36	100	100	46	9	26
Salisbury, MD-DE	47	85	51	23	28	51	38	57
Baltimore-Columbia-Towson, MD	44	44	53	20	16	54	100	20
Tallahassee, FL	43	80	51	26	19	85	13	27
Valdosta, GA	43	80	51	26	19	85	13	27
Alexandria, LA	39	59	40	23	8	63	19	58
Hattiesburg, MS	38	51	47	50	5	100	4	9
Montgomery, AL	37	73	47	19	14	63	14	32
Virginia Beach-Norfolk-Newport News, VA-NC	36	49	56	24	14	56	22	35
Auburn-Opelika, AL	36	47	62	24	8	71	17	23
Columbus, GA-AL	36	47	62	24	8	71	17	23
Augusta-Richmond County, GA-SC	36	32	18	60	35	44	28	34
Greenville, NC	35	53	58	25	30	34	19	27
Jacksonville, NC	35	53	58	25	30	34	19	27
Memphis, TN-MS-AR	35	34	62	34	19	54	15	28
Columbia, MO	35	46	53	44	51	24	8	18
Jefferson City, MO	35	46	53	44	51	24	8	18
Baton Rouge, LA	34	27	53	30	13	85	11	19
Monroe, LA	34	41	44	18	16	71	20	25
Canton-Massillon, OH	33	29	22	22	14	37	6	100
Mansfield, OH	33	29	22	22	14	37	6	100
Cleveland-Elyria, OH	33	29	22	22	14	37	6	100
Akron, OH	33	29	22	22	14	37	6	100
Salinas, CA	32	100	44	11	4	39	9	19
Santa Cruz-Watsonville, CA	32	100	44	11	4	39	9	19
Cape Girardeau, MO-IL	32	27	67	30	41	34	13	13
Savannah, GA	32	59	36	18	14	63	13	21
Richmond, VA	32	47	64	22	10	46	13	20
Myrtle Beach-Conway-North Myrtle Beach, SC-NC	32	34	44	34	23	37	23	26
Florence, SC	32	34	44	34	23	37	23	26
Beaumont-Port Arthur, TX	32	42	44	18	9	59	9	39
Springfield, MA	31	51	71	14	15	32	13	19
Albany, GA	31	36	67	19	1	78	6	6
Yuma, AZ	30	68	18	7	20	34	18	47
El Centro, CA	30	68	18	7	20	34	18	47
Columbia, SC	30	39	31	28	15	56	17	24
Sumter, SC	30	39	31	28	15	56	17	24
Youngstown-Warren-Boardman, OH-PA	30	31	36	19	9	39	9	65
Burlington, NC	29	34	47	18	15	49	15	27
	29	34		18	15			27
Greensboro-High Point, NC			47			49	15	
Winston-Salem, NC	29	34	47	18	15	49	15	27
Rockford, IL	29	46	47	18	34	34	12	14
Gadsden, AL	29	32	49	23	6	61	7	24
Fuscaloosa, AL	29	32	49	23	6	61	7	24
Birmingham-Hoover, AL	29	32	49	23	6	61	7	24
Anniston-Oxford-Jacksonville, AL	29	32	49	23	6	61	7	24
Lake Charles, LA	29	10	69	16	14	46	0	46
Wilmington, NC	29	46	42	19	9	46	20	19
Charleston-North Charleston, SC	29	29	60	27	11	54	13	6
Jackson, TN	28	25	40	27	19	22	12	54
Pine Bluff, AR	28	44	51	22	11	37	7	25
Little Rock-North Little Rock-Conway, AR	28	44	51	22	11	37	, 7	25
Hot Springs, AR	28	44	51	22	11	37	7	25
Minneapolis-St. Paul-Bloomington, MN-WI	28	29	100	11	11	24	6	15
St. Cloud, MN	28	29	100	11	11	24	6	15
New Orleans-Metairie, LA	28	29	42	26	16	24 59	12	18
Houma-Thibodaux, LA	28 28	24 24	42 42	26 26	16	59 59	12 12	18
Fayetteville, NC	28	39	40	19	13	51	8	23
Rocky Mount, NC	28	39	40	19	13	51	8	23
Goldsboro, NC	28	39	40	19	13	51	8	23
Durham-Chapel Hill, NC	28	39	40	19	13	51	8	23
Raleigh, NC	28	39	40	19	13	51	8	23
Lafayette, LA	28	32	40	18	10	51	8	33
Nashville-Davidson–Murfreesboro–Franklin, TN	27	25	36	22	16	59	13	22
Clarksville, TN-KY	27	25	36	22	16	59	13	22
Hickory-Lenoir-Morganton, NC	27	29	40	18	11	61	12	21
Charlotte-Concord-Gastonia, NC-SC	27	29	40	18	11	61	12	21
Orlando-Kissimmee-Sanford, FL	27	22	40	18	11	71	12	17
Deltona-Daytona Beach-Ormond Beach, FL	27	22	40	19	11	71	12	17
								1/

Table A.1 (Continued)

	Avg	Police	Black lives	Michael	Ferguson	Trayvon	Freddie	Tamir
Metro area	index	brutality	matter	Brown	unrest	Martin	Gray	Rice
Ocala, FL	27	22	40	19	11	71	12	17
Decatur, IL	27	39	53	25	18	34	7	15
Danville, IL	27	39	53	25	18	34	7	15
Champaign-Urbana, IL	27	39	53	25	18	34	7	15
Springfield, IL	27	39	53	25	18	34	7	15
Atlanta-Sandy Springs-Roswell, GA	27	22	44	22	11	59	9	23
Athens-Clarke County, GA	27	22	44	22	11	59	9	23
Gainesville, GA	27	22	44	22	11	59 59	9	23
Rome, GA Panama City, FL	27 27	22 32	44 27	22 20	11 24	59 54	9 7	23 26
Texarkana, TX-AR	27	36	73	8	24 10	46	3	20 13
Shreveport-Bossier City, LA	27	36	73	8	10	40	3	13
Dover, DE	27	39	31	20	13	49	17	18
Trenton, NJ	27	39	31	20	13	49	17	18
Reading, PA	27	39	31	20	13	49	17	18
Atlantic City-Hammonton, NJ	27	39	31	20	13	49	17	18
Ocean City, NJ	27	39	31	20	13	49	17	18
Philadelphia-Camden-Wilmington, PA-NJ-DE-MD	27	39	31	20	13	49	17	18
Allentown-Bethlehem-Easton, PA-NJ	27	39	31	20	13	49	17	18
Vineland-Bridgeton, NJ	27	39	31	20	13	49	17	18
Wheeling, WV-OH	27	66	13	8	20	41	16	21
Kingston, NY	26	41	38	20	10	39	12	23
New York-Newark-Jersey City, NY-NJ-PA	26	41	38	20	10	39	12	23
Bridgeport-Stamford-Norwalk, CT	26	41	38	20	10	39	12	23
Madera, CA	26	44	18	28	21	24	26	20
Visalia-Porterville, CA	26	44	18	28	21	24	26	20
Fresno, CA	26	44	18	28	21	24	26	20
Hanford-Corcoran, CA	26	44	18	28	21	24	26	20
Merced, CA	26	44	18	28	21	24	26	20
Topeka, KS	26	19	47	33	25	29	6	22
Manhattan, KS	26	19	47	33	25	29	6	22
Toledo, OH	26	42	42	15	14	27	11	29
Syracuse, NY	26	44	53	16	14	29	6	17
Ithaca, NY	26	44	53	16	14	29	6	17
Warner Robins, GA	25	36	33	17	1	54	15	22
Macon-Bibb County, GA	25	36	33	17	1	54	15	22
Utica-Rome, NY	25	46	33	16	14	32	13	24
Crestview-Fort Walton Beach-Destin, FL	25	17	44	30	18	46	7	15
Mobile, AL	25	17	44	30	18	46	7	15
Pensacola-Ferry Pass-Brent, FL Providence-Warwick, RI-MA	25	17 49	44	30 19	18 10	46	7	15 15
Bakersfield, CA	25 25	49	47 49	19	8	24 39	11 4	15
Jackson, MS	25	36	49	10	8 11	54	5	7
Winchester, VA-WV	25	30	33	11	10	39	23	19
Cumberland, MD-WV	25	31	33	18	10	39	23	19
Washington-Arlington-Alexandria, DC-VA-MD-	25	31	33	18	10	39	23	19
WV	20	51	55	10	10		20	17
Hagerstown-Martinsburg, MD-WV	25	31	33	18	10	39	23	19
Joplin, MO	25	12	49	18	29	22	0	43
Dothan, AL	25	32	33	19	11	44	14	19
Buffalo-Cheektowaga-Niagara Falls, NY	25	51	31	17	13	27	13	21
Monroe, MI	24	34	38	16	11	41	7	22
Detroit-Warren-Dearborn, MI	24	34	38	16	11	41	7	22
Ann Arbor, MI	24	34	38	16	11	41	7	22
Kankakee, IL	24	34	40	20	11	41	7	15
Michigan City-La Porte, IN	24	34	40	20	11	41	7	15
Chicago-Naperville-Elgin, IL-IN-WI	24	34	40	20	11	41	7	15
Gainesville, FL	24	25	22	15	11	61	12	22
Naples-Immokalee-Marco Island, FL	24	22	27	20	16	54	8	18
Cape Coral-Fort Myers, FL	24	22	27	20	16	54	8	18
Punta Gorda, FL	24	22	27	20	16	54	8	18
Columbus, IN	24	41	27	20	13	34	9	21
Indianapolis-Carmel-Anderson, IN	24	41	27	20	13	34	9	21
Bloomington, IN	24	41	27	20	13	34	9	21
Kokomo, IN	24	41	27	20	13	34	9	21
Muncie, IN	24	41	27	20	13	34	9	21
Reno, NV	23	46	42	15	18	20	16	8
Carson City, NV	23	46	42	15	18	20	16	8
Decatur, AL	23	15	38	28	15	37	6	24
Florence-Muscle Shoals, AL	23	15	38	28	15	37	6	24
Huntsville, AL	23	15	38	28	15	37	6	24

Table A.1 (Continued)

	Avg	Police	Black lives	Michael	Ferguson	Trayvon	Freddie	Tamir
Metro area	index	brutality	matter	Brown	unrest	Martin	Gray	Rice
Charlottesville, VA	23	19	40	20	16	29	16	22
Columbus, OH	23	27	33	19	9	32	9	33
Cincinnati, OH-KY-IN	23	25	44	20	11	27	11	23
Fort Smith, AR-OK	23	27	64	13	9	29	4	16
Fayetteville-Springdale-Rogers, AR-MO	23	27	64	13	9	29	4	16
Oklahoma City, OK	23	24	27	23	20	32	23	13
Gulfport-Biloxi-Pascagoula, MS	23	27	47	18	5	46	1	15
Asheville, NC	23	31	40	16	6	37	8	21
Greenville-Anderson-Mauldin, SC	23	31	40	16	6	37	8	21
Spartanburg, SC	23	31	40	16	6	37	8	21
Lansing-East Lansing, MI Jackson, MI	23 23	47 47	29 29	17 17	8 8	34 34	6 6	17 17
Brunswick, GA	23 22	17	29	17	8 10	68	9	17
Jacksonville, FL	22	17	20	16	10	68	9	17
Springfield, MO	22	32	31	10	31	15	13	16
Houston-The Woodlands-Sugar Land, TX	22	27	29	22	8	39	11	21
Springfield, OH	22	36	36	13	11	37	5	19
Dayton, OH	22	36	36	13	11	37	5	19
Harrisonburg, VA	22	37	7	23	25	29	16	19
Lincoln, NE	22	41	, 49	11	18	12	9	15
Rochester, NY	22	47	40	11	13	22	6	15
Bloomington, IL	22	37	36	19	14	20	9	19
Peoria, IL	22	37	36	19	14	20	9	19
Chattanooga, TN-GA	22	37	40	10	8	34	9	15
Dalton, GA	22	37	40	10	8	34	9	15
Cleveland, TN	22	37	40	10	8	34	9	15
Burlington-South Burlington, VT	22	25	67	10	9	15	5	21
Tucson, AZ	22	34	33	14	13	22	11	25
Bremerton-Silverdale, WA	22	29	42	14	16	24	7	18
Mount Vernon-Anacortes, WA	22	29	42	14	16	24	7	18
Olympia-Tumwater, WA	22	29	42	14	16	24	7	18
Wenatchee, WA	22	29	42	14	16	24	7	18
Bellingham, WA	22	29	42	14	16	24	7	18
Seattle-Tacoma-Bellevue, WA	22	29	42	14	16	24	7	18
Saginaw, MI	21	44	18	15	3	51	7	13
Flint, MI	21	44	18	15	3	51	7	13
Bay City, MI	21	44	18	15	3	51	7	13
Louisville/Jefferson County, KY-IN	21 21	27 27	20 20	27 27	19 19	27 27	15 15	15 15
Elizabethtown-Fort Knox, KY Sheboygan, WI	21 21	39	33	13	19	32	15 6	13
Racine, WI	21	39	33	13	14	32	6	13
Milwaukee-Waukesha-West Allis, WI	21	39	33	13	14	32	6	13
State College, PA	21	29	40	11	14	29	6	19
Johnstown, PA	21	29	40	11	14	29	6	19
Altoona, PA	21	29	40	11	14	29	6	19
Muskegon, MI	21	36	38	14	11	29	4	16
Grand Rapids-Wyoming, MI	21	36	38	14	11	29	4	16
Kalamazoo-Portage, MI	21	36	38	14	11	29	4	16
Battle Creek, MI	21	36	38	14	11	29	4	16
Lubbock, TX	21	39	36	16	13	24	6	14
Omaha-Council Bluffs, NE-IA	21	22	40	16	10	27	14	18
Erie, PA	21	34	47	18	5	20	9	13
Las Vegas-Henderson-Paradise, NV	21	25	40	15	9	34	8	14
Hartford-West Hartford-East Hartford, CT	21	37	36	11	8	29	5	19
Norwich-New London, CT	21	37	36	11	8	29	5	19
New Haven-Milford, CT	21	37	36	11	8	29	5	19
Miami-Fort Lauderdale-West Palm Beach, FL	21	22	13	11	3	73	5	17
Riverside-San Bernardino-Ontario, CA	20	37	27	17	9	37	5	12
Los Angeles-Long Beach-Anaheim, CA	20	37	27	17	9	37	5	12
Oxnard-Thousand Oaks-Ventura, CA	20	37	27	17	9	37	5	12
Kansas City, MO-KS	20	20	13	26	25	37	4	17
Lawrence, KS	20	20	13	26	25	37	4	17
Tampa-St. Petersburg-Clearwater, FL	20	17	33	11	9	49	12	12
Lakeland-Winter Haven, FL	20	17	33	11	9	49	12	12
North Port-Sarasota-Bradenton, FL	20	17	33	11	9	49	12	12
Kennewick-Richland, WA	20	44	38	9	21	10	7	13
Yakima, WA	20	44	38	9	21	10	7	13
Waco, TX College Station Prior TV	20	29	36 36	18	11	29	8	9 9
College Station-Bryan, TX								ч
Villoon Tomplo TV	20	29		18	11	29	8	
Killeen-Temple, TX Eugene, OR	20 20 20	29 29 32	36 44	18 18 9	11 11 19	29 29 20	8 3	9 14

Table A.1 (Continued)

	Avg	Police	Black lives	Michael	Ferguson	Trayvon	Freddie	Tamir
Metro area	index	brutality	matter	Brown	unrest	Martin	Gray	Rice
Lancaster, PA	20	27	24	13	18	27	18	13
Lebanon, PA	20	27	24	13	18	27	18	13
Harrisburg-Carlisle, PA	20	27	24	13	18	27	18	13
York-Hanover, PA	20	27	24	13	18	27	18	13
San Antonio-New Braunfels, TX	20	20	24	11	8	63	6	5
Madison, WI Janesville-Beloit, WI	20 20	32 32	44 44	11 11	10 10	17 17	6 6	17 17
San Diego-Carlsbad, CA	20	36	18	18	10	29	8	17
Boston-Cambridge-Newton, MA-NH	20	27	40	15	9	22	8	16
Manchester-Nashua, NH	20	27	40	15	9	22	8	16
Worcester, MA-CT	20	27	40	15	9	22	8	16
Barnstable Town, MA	20	27	40	15	9	22	8	16
Austin-Round Rock, TX	19	27	24	14	11	37	8	15
Colorado Springs, CO	19	34	27	16	14	27	8	11
Pueblo, CO	19	34	27	16	14	27	8	11
Morgantown, WV	19	29	22	14	11	27	13	19
Pittsburgh, PA	19	29	22	14	11	27	13	19
Coeur dAlene, ID	19	37	33	14	11	22	6	9
Lewiston, ID-WA	19	37	33	14	11	22	6	9
Spokane-Spokane Valley, WA	19 19	37 25	33 22	14	11 16	22 24	6 6	9 11
Wichita, KS Lawton, OK	19 19	25 24	22 40	27 13	16 14	24 22	6 13	11 7
Wichita Falls, TX	19	24	40	13	14	22	13	7
Knoxville, TN	19	20	16	15	16	39	7	18
Morristown, TN	19	20	16	15	16	39	7	18
Greeley, CO	19	37	24	16	15	22	5	11
Denver-Aurora-Lakewood, CO	19	37	24	16	15	22	5	11
Fort Collins, CO	19	37	24	16	15	22	5	11
Boulder, CO	19	37	24	16	15	22	5	11
Kingsport-Bristol-Bristol, TN-VA	19	29	56	8	10	20	8	0
Johnson City, TN	19	29	56	8	10	20	8	0
Davenport-Moline-Rock Island, IA-IL	18	34	20	17	21	20	8	7
Longview, TX	18	17	27	16	14	24	13	17
Tyler, TX	18	17	27	16	14	24	13	17
Fort Wayne, IN	18 18	39	22 31	14 13	8 9	22 20	7 9	15 7
Williamsport, PA Scranton–Wilkes-Barre–Hazleton, PA	18	37 37	31	13	9	20	9	7
Pittsfield, MA	18	44	20	10	8	20	7	17
Glens Falls, NY	18	44	20	10	8	20	7	17
Albany-Schenectady-Troy, NY	18	44	20	10	8	20	7	17
Binghamton, NY	18	34	29	10	3	27	7	15
Bangor, ME	17	32	29	11	13	5	5	27
Prescott, AZ	17	32	22	13	9	24	6	16
Lake Havasu City-Kingman, AZ	17	32	22	13	9	24	6	16
Phoenix-Mesa-Scottsdale, AZ	17	32	22	13	9	24	6	16
Flagstaff, AZ	17	32	22	13	9	24	6	16
Modesto, CA	17	32	18	13	13	27	5	15
Stockton-Lodi, CA	17	32	18	13	13	27	5	15
Sacramento–Roseville–Arden-Arcade, CA	17	32	18	13	13	27	5	15
Vallejo-Fairfield, CA	17	32	18	13	13	27	5	15
Yuba City, CA	17	32	18	13	13	27	5	15
Medford, OR	17 17	34 61	44	2 6	20	15	3	3 7
Las Cruces, NM El Paso, TX	17	61	18 18	6	6 6	17 17	6 6	7
Amarillo, TX	17	41	27	9	13	17	7	7
Iowa City, IA	17	29	27	15	10	24	8	7
Waterloo-Cedar Falls, IA	17	29	27	15	10	24	8	7
Cedar Rapids, IA	17	29	27	15	10	24	8	7
Dubuque, IA	17	29	27	15	10	24	8	7
Chico, CA	17	24	27	7	6	29	8	19
Redding, CA	17	24	27	7	6	29	8	19
Dallas-Fort Worth-Arlington, TX	17	20	24	14	8	37	5	12
Urban Honolulu, HI	17	29	13	24	14	15	5	18
Corpus Christi, TX	17	47	7	13	18	22	3	7
Wausau, WI	17	27	29	9	13	12	8	18
South Bend-Mishawaka, IN-MI	17	36	16	10	8	34	1	12
Niles-Benton Harbor, MI	17	36	16	10	8	34	1	12
Elkhart-Goshen, IN	17	36	16	10	8	34	1	12
Lexington-Fayette, KY	16 16	24	27	11	10	24	8	11 12
Lewiston-Auburn, ME Portland-South Portland, ME	16 16	36 36	33	11	5	10 10	8	
FOLUMIC-SOUTH FOLUMO, ME	16	36	33	11	5	10	8	12

Table A.1 (Continued)

	Avg	Police	Black lives	Michael	Ferguson	Trayvon	Freddie	Tamir
Metro area	index	brutality	matter	Brown	unrest	Martin	Gray	Rice
Rochester, MN	16	34	53	3	5	10	1	8
Grand Forks, ND-MN	16	8	60	11	4	12	3	16
Fargo, ND-MN	16	8	60	11	4	12	3	16
La Crosse-Onalaska, WI-MN	16	29	33	9	18	15	5	5
Eau Claire, WI	16	29	33	9	18	15	5	5
Santa Maria-Santa Barbara, CA	16	27	27	15	10	20	8	6
San Luis Obispo-Paso Robles-Arroyo Grande, CA	16	27	27	15	10	20	8	6
Missoula, MT	16	17	36	15	15	20	5	4
San Francisco-Oakland-Hayward, CA	16	29	20	13	10	20	6	14
Santa Rosa, CA	16	29	20	13	10	20	6	14
Napa, CA	16	29	20	13	10	20	6	14
San Jose-Sunnyvale-Santa Clara, CA	16	29	20	13	10	20	6	14
Huntington-Ashland, WV-KY-OH	16	31	24	11	11	20	7	6
Charleston, WV	16	31	24	11	11	20	7	6
Sebastian-Vero Beach, FL	16	19	2	11	6	61	3	7
Port St. Lucie, FL	16	19	2	11	6	61	3	7
Farmington, NM	16	41	24	8	10	20	4	2
Santa Fe, NM	16	41	24	8	10	20	4	2
Albuquerque, NM	16	41	24	8	10	20	4	2
Bismarck, ND	15	24	20	3	24	15	5	14
Owensboro, KY	15	19	13	14	21	22	13	1
Evansville, IN-KY	15	19	13	14	21	22	13	1
Green Bay, WI	14	31	20	10	10	17	5	7
Fond du Lac, WI	14	31	20	10	10	17	5	7
Appleton, WI	14	31	20	10	10	17	5	7
Ames, IA	14	24	18	10	13	17	6	12
Des Moines-West Des Moines, IA	14	24	18	10	13	17	6	12
Midland, TX	14	17	29	10	4	12	9	16
Odessa, TX	14	17	29	10	4	12	9	16
Boise City, ID	13	24	31	8	5	10	5	9
St. George, UT	12	19	22	8	9	15	7	4
Logan, UT-ID	12	19	22	8	9	15	, 7	4
Salt Lake City, UT	12	19	22	8	9	15	, 7	4
Provo-Orem, UT	12	19	22	8	9	15	7	4
Ogden-Clearfield, UT	12	19	22	8	9	15	7	4
Duluth, MN-WI	12	7	42	3	8	17	2	2
Tulsa, OK	11	10	13	11	8	15	7	12
Anchorage, AK	11	2	18	3	8	24	4	16
Sioux Falls, SD	11	20	27	3 7	3	24 7	4	6
Brownsville-Harlingen, TX	9	31	4	8	0	10	1	8
McAllen-Edinburg-Mission, TX	9	31	4	8	0	10	1	8
Salem, OR	8	12	4 16	8	6	7	2	8 13
Salem, OK Portland-Vancouver-Hillsboro, OR-WA	8	12 12	16	2	6	7	2	13
	8	12		2		7	2	13
Longview, WA		12	16 0	2	6 3	7	2	13
Roanoke, VA	$\frac{2}{2}$							
Lynchburg, VA Blocksburg Christiansburg Dadford MA	2 2	0	0	0	3	7 7	4	3
Blacksburg-Christiansburg-Radford, VA	2	0	0	0	3	/	4	3

Note: Variation in Google Trends metrics is available at the Designated Market Area level, which we translate to the metro area level. The table includes metro areas used in the main analysis that do not contain missing values for any of the metrics listed. For purposes of the table only, the indexes were normalized to range from 0 to 100. Data from Google Trends was extracted on 3/7/17.

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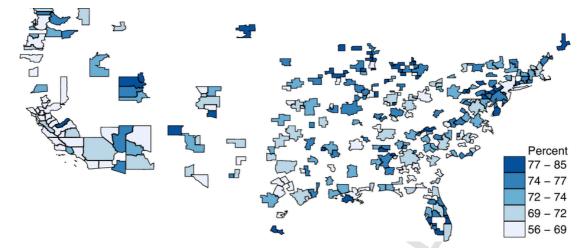


Fig. 2. Homeownership Rates of Movers in New Location by Metro Area, 2011. Note: Only CBSAs used in the analysis are shown, which represent 82.7% of the population in 2010. Data come from the ACS for household that moved in the last year.

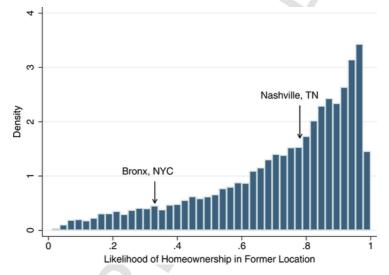


Fig. 3. Homeowner Averages in Origin Location by Demographic Groups. Note: Homeownership is former location (Migration PUMA) is derived by taking the average homeownership of households with similar demographics in an observations former location. Demographics include the household head's age bin (18–34, 35–54, 55+), income quartile, marital status, and race. The sample is restricted to observations where there are at least 100 household heads in their former location the fall into the same demographic.

Table A.2

Robustness check: alternative quantiles.

	Full	Full			Movers	Movers	
	(1)	(2)	(3)	(4)	(5)	(6)	
$Black_i \times$	- 0.017		026**		025*		
Aggregate Index GT Top Tercile _i	(0.011)		(0.011)		(0.015)		
$Black_i \times$		023*		061***		073***	
Aggregate Index GT Top Quintile _j Obs.	3,709,054	(0.012) 1,684,649	635,352	(0.018) 303,756	99,473	(0.018) 44,273	

Note: Dependent variable is homeownership. The full sample includes white or black headed households from the ACS in years 2005–2011. The sample is restricted to households that live in areas that fall into the upper and lower quantiles (either tercile or quintile respectively) for the Aggregate Index of racial climate. Controls included but not reported include age, race, education, marital status, child present, metro level characteristics, interactions of metro characteristics and race, CBSA fixed effects, and year fixed effects. Aggregate Index for Racial Climate uses the following Google Search Terms/Topics: Police Brutality, Black Lives Matter, Shooting of Michael Brown, Ferguson Unrest, Trayvon Martin, Death of Freddie Gray and Shooting of Tamir Rice. Standard errors are clustered at the DMA level and are shown in parentheses *** p < .01, ** p < .05, * p < .1.

Table A.3

Influence of metro controls by race interactions: Gelbach decomposition.

Racial climate index point estimate:	Young	Across metro movers
Without Interactions	- 0.018	- 0.019
	(0.028)	(0.024)
With Interactions	-	056***
	.042***	
	(0.024)	(0.016)
Gelbach Decomposition		
$Black_i \times$		
Percent Black _i	-	045**
,	.038***	
	(0.012)	(0.018)
Log Population _i (100)	- 0.005	- 0.007
	(0.007)	(0.006)
Crime Rates _{j, t} (Z-score)	- 0.005	0.001
<i></i>	(0.003)	(0.004)
FHFA HPI _{<i>j</i>, <i>t</i>} (100)	- 0.001	- 0.002
	(0.002)	(0.003)
FMR _{j, t}	0.023	0.024
	(0.021)	(0.023)
Income-to-poverty ratio _j	0.002	- 0.009
	(0.009)	(0.014)
Share manufacturing _{j, t}	0.000	0.001
	(0.002)	(0.002)

Note: Numbers reported reflect the influence of each covariate in the change of the Race Relations coefficient from the model without metro controls by race interactions to the full specification using a technique described in Glebach (2016). Essentially, we estimate omitted variable bias on the coefficient of interest (in this case Racial Climate) from the exclusion of each control one at a time. The sum of an individual column in the lower panel will fully describe the change in the point estimate across the two models (shown in the top panel). Standard errors are shown in parentheses *** p < .01, ** p < .05, * p < .1.

Table A.4

Determinants of homeownership.

	Coeff.	R^2 if Omitted
Black	165***	0.181
	(0.000)	
Age 35–54	.211***	0.145
	(0.000)	
Age 55+	.328***	
	(0.000)	
Married	.140***	0.174
	(0.000)	
Has Child	.031***	0.196
	(0.000)	
High School or less	003***	0.196
	(0.000)	
Household Income 2nd Quartile	$.128^{***}$	0.163
	(0.000)	
Household Income 3rd Quartile	.198***	
	(0.000)	
Household Income 4th Quartile	.234***	
	(0.000)	
Obs.	11,104,464	
R-Squared	0.196	

Note: Sample consists of non-movers in the ACS from 2005 to 2015. Standard errors are shown in parentheses *** p < .01, ** p < .05, * p < .1.

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Table A.5

DMA	CBSA
Abilene-Sweetwater TX	Brownwood, TX
Abilene-Sweetwater TX	Sweetwater, TX
Abilene-Sweetwater TX	Snyder, TX
Abilene-Sweetwater TX	Abilene, TX
Albany GA	Albany, GA
Albany GA	Cordele, GA
Albany GA	Fitzgerald, GA
Albany GA	Tifton, GA
Albany GA	Douglas, GA
Albany GA	Moultrie, GA
Albany-Schenectady-Troy NY	Bennington, VT
Albany-Schenectady-Troy NY	Albany-Schenectady-Troy, NY
Albany-Schenectady-Troy NY	Gloversville, NY
Albany-Schenectady-Troy NY	Pittsfield, MA
Albany-Schenectady-Troy NY	Glens Falls, NY
Albany-Schenectady-Troy NY	Hudson, NY
Albany-Schenectady-Troy NY	Amsterdam, NY
Albuquerque-Santa Fe NM	Grants, NM
Albuquerque-Santa Fe NM	Farmington, NM
Albuquerque-Santa Fe NM	Los Alamos, NM
Albuquerque-Santa Fe NM	Durango, CO
Albuquerque-Santa Fe NM	Albuquerque, NM
Albuquerque-Santa Fe NM	Silver City, NM
Albuquerque-Santa Fe NM	Gallup, NM
Albuquerque-Santa Fe NM	Alamogordo, NM
Albuquerque-Santa Fe NM	Carlsbad-Artesia, NM
Albuquerque-Santa Fe NM	Santa Fe, NM
Albuquerque-Santa Fe NM	Hobbs, NM
Albuquerque-Santa Fe NM	Taos, NM
Albuquerque-Santa Fe NM	Las Vegas, NM
Albuquerque-Santa Fe NM	Deming, NM
Albuquerque-Santa Fe NM	Espaola, NM
Albuquerque-Santa Fe NM	Roswell, NM
Alexandria LA	Fort Polk South, LA
Alexandria LA	Alexandria, LA
Alpena MI Amarillo TX	Alpena, MI
Amarillo TX	Portales, NM Dumas, TX
Amarillo TX	,
Amarillo TX	Amarillo, TX Hereford, TX
Amarillo TX	Borger, TX
Amarillo TX	Pampa, TX
Amarillo TX	Clovis, NM
Amarillo TX	Guymon, OK
Anchorage AK	Anchorage, AK
Atlanta GA	Athens-Clarke County, GA
Atlanta GA	Atlanta-Sandy Springs-Roswell,
	GA
Atlanta GA	Gainesville, GA
Atlanta GA	Cedartown, GA
Atlanta GA	Thomaston, GA
Atlanta GA	Rome, GA
Atlanta GA	Calhoun, GA
Atlanta GA	LaGrange, GA
Atlanta GA	Cornelia, GA
Atlanta GA	Jefferson, GA
Augusta GA	Augusta-Richmond County, GA-
	SC
Augusta GA	Augusta-Richmond County, GA-SC
Austin TY	
Austin TX	Austin-Round Rock, TX
Austin TX Bakersfield CA	Fredericksburg, TX Bakersfield, CA
Bakersfield CA Reltimore MD	Bakersfield, CA Baltimoro Columbia Touron, MI
Baltimore MD	Baltimore-Columbia-Towson, MI
Baltimore MD	Easton, MD Dhiladalphia Comdon
Baltimore MD	Philadelphia-Camden- Wilmington, PA-NJ-DE-MD
Bangor ME	Bangor, ME
Baton Rouge LA	Baton Rouge, LA
Baton Rouge LA	McComb, MS
Baton Rouge LA	Morgan City, LA
Beaumont-Port Arthur TX	Beaumont-Port Arthur, TX
Bend OR	Bend-Redmond, OR

Table A.5 (Continued)

DMA

Biloxi-Gulfport MS Binghamton NY Birmingham AL Birmingham AL Birmingham AL Birmingham AL Birmingham AL Birmingham AL Bluefield-Beckley-Oak Hill WV Bluefield-Beckley-Oak Hill WV Bluefield-Beckley-Oak Hill WV Boise ID Boise ID Boise ID Boise ID Boise ID Boston MA-Manchester NH Bowling Green KY Bowling Green KY Buffalo NY Buffalo NY Buffalo NY Buffalo NY Buffalo NY Buffalo NY Burlington VT-Plattsburgh NY Butte-Bozeman Mount Butte-Bozeman Mount Butte-Bozeman Mount Casper-Riverton WY Casper-Riverton WY Cedar Rapids-Waterloo-Iowa City & Dubuque IA Champaign & Springfield-Decatur IL Charleston SC Charleston SC Charleston-Huntington WV Charleston-Huntington WV Charleston-Huntington WV Charleston-Huntington WV Charleston-Huntington WV

CBSA

Gulfport-Biloxi-Pascagoula, MS Binghamton, NY Tuscaloosa, AL Birmingham-Hoover, AL Anniston-Oxford-Jacksonville, AL Gadsden AL Talladega-Sylacauga, AL Cullman, AL Bluefield, WV-VA Bluefield, WV-VA Beckley, WV Boise City, ID Mountain Home, ID Hailey, ID Ontario, OR-ID Ontario, OR-ID Boston-Cambridge-Newton, MA-NH Manchester-Nashua, NH Barnstable Town, MA Worcester, MA-CT Laconia, NH Vinevard Haven, MA Boston-Cambridge-Newton, MA-NH Keene, NH Concord, NH Bowling Green, KY Glasgow, KY Olean, NY Jamestown-Dunkirk-Fredonia, NY Rochester, NY Buffalo-Cheektowaga-Niagara Falls, NY Batavia, NY Bradford, PA Berlin, NH-VT Burlington-South Burlington, VT Claremont-Lebanon, NH-VT Plattsburgh, NY Rutland, VT Claremont-Lebanon NH-VT Malone, NY Barre, VT Bozeman, MT Helena, MT Butte-Silver Bow, MT Riverton, WY Casper, WY Waterloo-Cedar Falls, IA Iowa City, IA

Dubuque, IA

Cedar Rapids, IA

Effingham, IL Champaign-Urbana, IL Lincoln, IL Decatur, IL Charleston-Mattoon, IL Taylorville, IL Springfield, IL Jacksonville, IL Danville, IL Bloomington, IL Georgetown, SC Charleston-North Charleston, SC Jackson, OH Logan, WV Charleston, WV Athens, OH Huntington-Ashland, WV-KY-OH

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Fable A.5 (Continued)	
DMA	CBSA
Charleston-Huntington WV	Huntington-Ashland, WV-KY-OH
Charleston-Huntington WV	Huntington-Ashland, WV-KY-OH
Charleston-Huntington WV	Portsmouth, OH
Charleston-Huntington WV	Parkersburg-Vienna, WV
Charleston-Huntington WV Charleston-Huntington WV	Point Pleasant, WV-OH Point Pleasant, WV-OH
Charlotte NC	Hickory-Lenoir-Morganton, NC
Charlotte NC	Charlotte-Concord-Gastonia, NC-SC
Charlotte NC	Rockingham, NC
Charlotte NC Charlotte NC	Boone, NC Charlotte-Concord-Gastonia, NC-
Charlotte NC	SC Albemarle, NC
Charlotte NC	Shelby, NC
Charlottesville VA	Charlottesville, VA
Chattanooga TN	Chattanooga, TN-GA
Chattanooga TN	Summerville, GA
Chattanooga TN	Dalton, GA
Chattanooga TN Chattanooga TN	Cleveland, TN
Chattanooga TN Chattanooga TN	Chattanooga, TN-GA Dayton, TN
Chattanooga TN Chattanooga TN	Athens, TN
Cheyenne WY-Scottsbluff NE	Cheyenne, WY
Cheyenne WY-Scottsbluff NE	Scottsbluff, NE
Chicago IL	Chicago-Naperville-Elgin, IL-IN- WI
Chicago IL	Chicago-Naperville-Elgin, IL-IN-
Chicago IL	WI Michigan City-La Porte, IN
Chicago IL	Ottawa-Peru, IL
Chicago IL	Kankakee, IL
Chico-Redding CA	Redding, CA
Chico-Redding CA	Chico, CA
Chico-Redding CA	Red Bluff, CA
Cincinnati OH	Cincinnati, OH-KY-IN
Cincinnati OH Cincinnati OH	Cincinnati, OH-KY-IN
Cincinnati OH	Cincinnati, OH-KY-IN Maysville, KY
Cincinnati OH	Wilmington, OH
Clarksburg-Weston WV	Clarksburg, WV
Clarksburg-Weston WV	Elkins, WV
Clarksburg-Weston WV	Fairmont, WV
Cleveland-Akron (Canton) OH	Ashtabula, OH
Cleveland-Akron (Canton) OH Cleveland-Akron (Canton) OH	Ashland, OH Cleveland-Elyria, OH
Cleveland-Akron (Canton) OH	Akron, OH
Cleveland-Akron (Canton) OH	Canton-Massillon, OH
Cleveland-Akron (Canton) OH	Mansfield, OH
Cleveland-Akron (Canton) OH	Norwalk, OH
Cleveland-Akron (Canton) OH Cleveland-Akron (Canton) OH	New Philadelphia-Dover, OH Wooster, OH
Cleveland-Akron (Canton) OH	Sandusky, OH
Colorado Springs-Pueblo CO	Caon City, CO
Colorado Springs-Pueblo CO	Colorado Springs, CO
Colorado Springs-Pueblo CO	Pueblo, CO
Columbia SC	Columbia, SC
Columbia SC Columbia SC	Newberry, SC
Columbia SC Columbia SC	Orangeburg, SC Sumter, SC
Columbia-Jefferson City MO	Jefferson City, MO
Columbia-Jefferson City MO	Mexico, MO
Columbia-Jefferson City MO	Columbia, MO
Columbia-Jefferson City MO	Moberly, MO
Columbus GA	Columbus, GA-AL
Columbus GA	Auburn-Opelika, AL
Columbus GA Columbus GA	Columbus, GA-AL Americus, GA
Columbus GA Columbus GA	Valley, AL
Columbus OH	Columbus, OH
Columbus OH	Mount Vernon, OH
Columbus OH	Washington Court House, OH
Columbus OH	Bucyrus, OH
Columbus OH	Cambridge, OH

DMA Columbus OH Columbus OH Columbus OH Columbus-Tupelo-West Point MS Columbus-Tupelo-West Point MS Columbus-Tupelo-West Point MS Corpus Christi TX Corpus Christi TX Corpus Christi TX Corpus Christi TX Dallas-Ft. Worth TX Dallas-Ft. Worth TX Dallas-Ft. Worth TX Dallas-Ft, Worth TX Dallas-Ft. Worth TX Davenport IA-Rock Island-Moline IL Davton OH Dayton OH Dayton OH Davton OH Dayton OH Dayton OH Dayton OH Davton OH Denver CO Des Moines-Ames IA Detroit MI Detroit MI Detroit MI Dothan AL

Dothan AL

Dothan AL

El Paso TX

El Paso TX

Elmira NY

Elmira NY

Erie PA

Erie PA

Erie PA

Eugene OR

Duluth MN-Superior WI

Duluth MN-Superior WI

CBSA Coshocton, OH Chillicothe, OH Marion, OH Columbus, MS Tupelo, MS Starkville, MS Corpus Christi, TX Kingsville, TX Alice, TX Beeville, TX Dallas-Fort Worth-Arlington, TX Mineral Wells, TX Paris, TX Corsicana, TX Stephenville, TX Gainesville, TX Athens, TX Sulphur Springs, TX Palestine, TX Davenport-Moline-Rock Island, IA-IL Galesburg, IL Ottawa-Peru, IL Muscatine, IA Burlington, IA-IL Burlington, IA-IL Davenport-Moline-Rock Island, IA-IL Sterling, IL Clinton, IA Dayton, OH Urbana, OH Springfield, OH Bellefontaine, OH Richmond, IN Sidney, OH Greenville, OH Celina, OH Fort Collins, CO Denver-Aurora-Lakewood, CO Sterling, CO Craig, CO Steamboat Springs, CO Gillette, WY Greeley, CO Glenwood Springs, CO Laramie, WY Breckenridge, CO Boulder, CO Fort Morgan, CO Edwards, CO Des Moines-West Des Moines, IA Boone, IA Ames, IA Marshalltown, IA Fort Dodge, IA Newton, IA Oskaloosa, IA Ann Arbor, MI Detroit-Warren-Dearborn, MI Monroe, MI Dothan, AL Enterprise, AL Ozark, AL Duluth, MN-WI Duluth, MN-WI El Paso, TX Las Cruces, NM Elmira, NY Corning, NY Meadville, PA Erie, PA Warren, PA Eugene, OR

Eugene OR	Roseburg, OR
Eugene OR	Coos Bay, OR
Eureka CA	Eureka-Arcata-Fortuna, CA
Eureka CA	Crescent City, CA
Evansville IN	Evansville, IN-KY
Evansville IN	Owensboro, KY
Evansville IN	Jasper, IN
Evansville IN	Evansville, IN-KY
Evansville IN	Madisonville, KY
Fairbanks AK	Fairbanks, AK
Fargo-Valley City ND	Wahpeton, ND-MN
Fargo-Valley City ND	Grand Forks, ND-MN
Fargo-Valley City ND	Fergus Falls, MN
Fargo-Valley City ND	Grand Forks, ND-MN
Fargo-Valley City ND	Wahpeton, ND-MN
Fargo-Valley City ND	Jamestown, ND
Fargo-Valley City ND	Fargo, ND-MN
Fargo-Valley City ND	Fargo, ND-MN
Flint-Saginaw-Bay City MI	Alma, MI
Flint-Saginaw-Bay City MI	Mount Pleasant, MI
Flint-Saginaw-Bay City MI	Flint, MI
Flint-Saginaw-Bay City MI	Saginaw, MI
Flint-Saginaw-Bay City MI	Midland, MI
Flint-Saginaw-Bay City MI	Owosso, MI
Flint-Saginaw-Bay City MI	Bay City, MI
Florence-Myrtle Beach SC	Laurinburg, NC
Florence-Myrtle Beach SC	Myrtle Beach-Conway-North
Plotence-wythe beach 3C	Myrtle Beach, SC-NC
Florence-Myrtle Beach SC	Florence, SC
Florence-Myrtle Beach SC	Lumberton, NC
Florence-Myrtle Beach SC	Bennettsville, SC
Fresno-Visalia CA	Merced, CA
Fresno-Visalia CA	Fresno, CA
Fresno-Visalia CA	Hanford-Corcoran, CA
Fresno-Visalia CA	Visalia-Porterville, CA
Fresno-Visalia CA	Madera, CA
Ft. Myers-Naples FL	Punta Gorda, FL
Ft. Myers-Naples FL	Arcadia, FL
Ft. Myers-Naples FL	Clewiston, FL
Ft. Myers-Naples FL	Cape Coral-Fort Myers, FL
Ft. Myers-Naples FL	Naples-Immokalee-Marco Island,
rt. Wyeis-Wapies rL	FL
Ft. Smith-Fayetteville-Springdale-Rogers	Fort Smith, AR-OK
AR	Fort Sillini, AR-OK
Ft. Smith-Fayetteville-Springdale-Rogers	Fayetteville-Springdale-Rogers,
AR	AR-MO
Ft. Smith-Fayetteville-Springdale-Rogers	Fort Smith, AR-OK
AR	Fort Sillini, AR-OK
	Fort Wayne, IN
Ft. Wayne IN Ft. Wayne IN	Wabash, IN
-	Decatur, IN
Ft. Wayne IN Ft. Wayne IN	
•	Auburn, IN
Ft. Wayne IN	Angola, IN Kondollyrillo, IN
Ft. Wayne IN	Kendallville, IN Van Wort, OH
Ft. Wayne IN	Van Wert, OH Huntington, IN
Ft. Wayne IN Gainesville FL	Huntington, IN
	Gainesville, FL Grand Junction, CO
Grand Junction-Montrose CO	,
Grand Junction-Montrose CO	Montrose, CO
Grand Rapids-Kalamazoo-Battle Creek MI	Coldwater, MI Crond Danida Wyoming, MI
Grand Rapids-Kalamazoo-Battle Creek MI	Grand Rapids-Wyoming, MI
Grand Rapids-Kalamazoo-Battle Creek MI Grand Banids-Kalamazoo-Battle Creek MI	Kalamazoo-Portage, MI Holland, MI

Holland, MI

Ionia, MI

Sturgis, MI

Muskegon, MI

Battle Creek, MI

Great Falls, MT

Manitowoc, WI

Shawano, WI

Green Bay, WI

Appleton, WI

Oshkosh-Neenah, WI

CBSA

Corvallis, OR

Roseburg, OR

Table A.5 (Continued)

DMA

Eugene OR

Eugene OR

Great Falls Mount

Green Bay-Appleton WI

Grand Rapids-Kalamazoo-Battle Creek MI

DMA Green Bay-Appleton WI Green Bay-Appleton WI Green Bay-Appleton WI Greensboro-High Point-Winston Salem NC Greenville-New Bern-Washington NC Greenville-Spartanburg SC-Asheville NC-Anderson SC Greenwood-Greenville MS Greenwood-Greenville MS Greenwood-Greenville MS Greenwood-Greenville MS Greenwood-Greenville MS Harlingen-Weslaco-Brownsville-McAllen ТХ Harlingen-Weslaco-Brownsville-McAllen TX Harlingen-Weslaco-Brownsville-McAllen ΤX Harlingen-Weslaco-Brownsville-McAllen TX Harrisburg-Lancaster-Lebanon-York PA Harrisburg-Lancaster-Lebanon-York PA Harrisburg-Lancaster-Lebanon-York PA Harrisburg-Lancaster-Lebanon-York PA Harrisburg-Lancaster-Lebanon-York PA Harrisburg-Lancaster-Lebanon-York PA Harrisonburg VA Harrisonburg VA Hartford & New Haven CT Hattiesburg-Laurel MS

Hattiesburg-Laurel MS

Helena Mount

Honolulu HI

Honolulu HI

Honolulu HI

Honolulu HI

Houston TX

Houston TX

Marinette, WI-MI Winston-Salem, NC Mount Airy, NC Greensboro-High Point, NC North Wilkesboro, NC Burlington, NC New Bern, NC Washington, NC Kill Devil Hills, NC Jacksonville, NC Greenville, NC Kinston, NC Morehead City, NC Cullowhee, NC Greenville-Anderson-Mauldin, SC Greenwood, SC Spartanburg, SC Marion, NC Forest City, NC Seneca, SC Toccoa, GA Asheville, NC Gaffney, SC Brevard, NC Grenada, MS Indianola, MS Greenwood, MS Greenville, MS Cleveland, MS Brownsville-Harlingen, TX Raymondville, TX Rio Grande City, TX McAllen-Edinburg-Mission, TX Harrisburg-Carlisle, PA Lewistown, PA Lancaster, PA York-Hanover, PA Gettysburg, PA Lebanon, PA Harrisonburg, VA Staunton-Waynesboro, VA Hartford-West Hartford-East Hartford, CT Torrington, CT New Haven-Milford, CT Worcester, MA-CT Norwich-New London, CT Laurel, MS Hattiesburg, MS Helena, MT Kahului-Wailuku-Lahaina, HI Hilo, HI Urban Honolulu, HI Kapaa, HI Houston-The Woodlands-Sugar Land, TX El Campo, TX

CBSA

Fond du Lac, WI

Marinette, WI-MI

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A.5 (Continued) Table

Fable A.5 (Continued)	
DMA	CBSA
Houston TX	Huntsville, TX
Houston TX	Bay City, TX
Houston TX	Port Lavaca, TX Brenham, TX
Houston TX Huntsville-Decatur (Florence) AL	Huntsville, AL
Huntsville-Decatur (Florence) AL	Florence-Muscle Shoals, AL
Huntsville-Decatur (Florence) AL	Albertville, AL
Huntsville-Decatur (Florence) AL	Scottsboro, AL
Huntsville-Decatur (Florence) AL	Decatur, AL
Idaho Falls-Pocatello ID Idaho Falls-Pocatello ID	Idaho Falls, ID Jackson, WY-ID
Idaho Falls-Pocatello ID	Pocatello, ID
Idaho Falls-Pocatello ID	Jackson, WY-ID
Idaho Falls-Pocatello ID	Rexburg, ID
Idaho Falls-Pocatello ID	Blackfoot, ID
Indianapolis IN	Peru, IN
Indianapolis IN	Marion, IN
Indianapolis IN Indianapolis IN	Greensburg, IN Indianapolis-Carmel-Anderson,
inchairapons inv	IN
Indianapolis IN	Lafayette-West Lafayette, IN
Indianapolis IN	New Castle, IN
Indianapolis IN	Connersville, IN
Indianapolis IN	Columbus, IN
Indianapolis IN	Crawfordsville, IN
Indianapolis IN Indianapolis IN	Bloomington, IN Bedford, IN
Indianapolis IN	Logansport, IN
Indianapolis IN	Muncie, IN
Indianapolis IN	Frankfort, IN
Indianapolis IN	Kokomo, IN
Jackson MS	Jackson, MS
Jackson MS Jackson MS	Natchez, MS-LA McComb, MS
Jackson MS	Vicksburg, MS
Jackson MS	Brookhaven, MS
Jackson TN	Jackson, TN
Jacksonville FL	Brunswick, GA
Jacksonville FL	Lake City, FL
Jacksonville FL	Jacksonville, FL
Jacksonville FL Jacksonville FL	St. Marys, GA Palatka, FL
Jacksonville FL	Waycross, GA
Johnstown-Altoona PA	State College, PA
Johnstown-Altoona PA	Altoona, PA
Johnstown-Altoona PA	Huntingdon, PA
Johnstown-Altoona PA Johnstown-Altoona PA	Somerset, PA
Johnstown-Altoona PA	Johnstown, PA DuBois, PA
Jonesboro AR	Jonesboro, AR
Jonesboro AR	Paragould, AR
Joplin MO-Pittsburg KS	Fayetteville-Springdale-Rogers,
	AR-MO
Joplin MO-Pittsburg KS Joplin MO-Pittsburg KS	Joplin, MO Parsons, KS
Joplin MO-Pittsburg KS	Pittsburg, KS
Joplin MO-Pittsburg KS	Miami, OK
Juneau AK	Juneau, AK
Kansas City MO	Kansas City, MO-KS
Kansas City MO	Marshall, MO
Kansas City MO	Lawrence, KS
Kansas City MO Kansas City MO	Kansas City, MO-KS Atchison, KS
Kansas City MO	Sedalia, MO
Kansas City MO	Warrensburg, MO
Kansas City MO	Ottawa, KS
Knoxville TN	Morristown, TN
Knoxville TN	Knoxville, TN
Knoxville TN Knoxville TN	Middlesborough, KY Sevierville, TN
Knoxville TN Knoxville TN	Crossville, TN
Knoxville TN	Newport, TN
La Crosse-Eau Claire WI	La Crosse-Onalaska, WI-MN
La Crosse-Eau Claire WI	Winona, MN
La Crosse-Eau Claire WI	Eau Claire, WI

DMA

La Crosse-Eau Claire WI La Crosse-Eau Claire WI Lafayette IN Lafayette LA Lafayette LA Lake Charles LA Lake Charles LA Lansing MI Lansing MI Lansing MI Laredo TX Laredo TX Las Vegas NV Las Vegas NV Lexington KY Lima OH Lima OH Lincoln & Hastings-Kearney NE Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR

Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR Little Rock-Pine Bluff AR Los Angeles CA

Los Angeles CA

Los Angeles CA

Louisville KY Louisville KY

Louisville KY Louisville KY

Louisville KY Louisville KY Louisville KY Louisville KY Lubbock TX Lubbock TX Lubbock TX Lubbock TX Macon GA Macon GA Macon GA Macon GA Madison WI Madison WI Madison WI Madison WI Mankato MN Mankato MN Marquette MI Marquette MI Marquette MI Marquette MI

CBSA

Menomonie, WI La Crosse-Onalaska, WI-MN Lafayette-West Lafayette, IN Lafayette, LA Opelousas, LA Lake Charles, LA DeRidder, LA Jackson, MI Lansing-East Lansing, MI Hillsdale, MI Zapata, TX Laredo, TX Las Vegas-Henderson-Paradise, NV Pahrump, NV Frankfort, KY Lexington-Fayette, KY Richmond-Berea, KY London, KY Danville, KY Mount Sterling, KY Somerset, KY Lima, OH Wapakoneta, OH Hastings, NE Grand Island, NE Kearney, NE Beatrice, NE Lincoln, NE Lexington, NE Russellville, AR Little Rock-North Little Rock-Conway, AR Arkadelphia, AR Pine Bluff, AR Batesville, AR Malvern, AR Searcy, AR Camden, AR Hot Springs, AR Oxnard-Thousand Oaks-Ventura, CA Los Angeles-Long Beach-Anaheim, CA Riverside-San Bernardino-Ontario, CA Campbellsville, KY Louisville/Jefferson County, KY-IN Madison, IN Louisville/Jefferson County, KY-IN Elizabethtown-Fort Knox, KY Bardstown, KY North Vernon, IN Seymour, IN Lubbock, TX Levelland, TX Plainview, TX Lamesa, TX Macon-Bibb County, GA Warner Robins, GA Milledgeville, GA Dublin, GA Madison, WI Baraboo, WI Janesville-Beloit, WI Platteville, WI New Ulm, MN Mankato-North Mankato, MN Escanaba, MI Iron Mountain, MI-WI Iron Mountain, MI-WI Houghton, MI

Table A.5 (Continued)	
DMA	CBS
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Memphis TN	Blyt
Memphis TN	Jon
Memphis TN	Cor
Memphis TN	Oxf
Memphis TN	Hele
Memphis TN	For
Memphis TN	Cla
Memphis TN	Jack
Meridian MS Miami-Ft. Lauderdale FI	Mer Key
Miami-Ft. Lauderdale FI	
Midilii I I. Eddderddie I I	Palı
Milwaukee WI	Rac
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Minot-Bismarck-Dickins	
Minot-Bismarck-Dickins	
Minot-Bismarck-Dickins	
Minot-Bismarck-Dickins Missoula Mount	Mis
Missoula Mount	Kali
Mobile AL-Pensacola (Ft	
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Mobile AL-Pensacola (Ft	
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Mobile AL-Pensacola (Ft	. Walton Beach) Dap
FL Monroe LA-El Dorado A	R Mor
Monroe LA-El Dorado A	
Monterey-Salinas CA	Sali
Monterey-Salinas CA	San
	CA
Monterey-Salinas CA	San
Montgomery (Selma) AL	
Montgomery (Selma) AI Montgomery (Selma) AI	
Montgomery (Selma) AL	
Nashville TN	Tull

Table A.5 (Continued)

SΑ rquette, MI nts Pass. OR math Falls, OR okings, OR lford, OR nphis, TN-MS-AR nphis, TN-MS-AR rsburg, TN nphis, TN-MS-AR theville, AR esboro, AR inth, MS ord, MS ena-West Helena, AR rest City, AR rksdale, MS kson, TN ridian. MS West, FL mi-Fort Lauderdale-West m Beach, FL ine, WI itewater-Elkhorn, WI waukee-Waukesha-West Allis, ver Dam, WI tertown-Fort Atkinson, WI boygan, WI cago-Naperville-Elgin, IL-INneapolis-St. Paulomington, MN-WI inerd, MN Wing, MN hester, MN shall, MN ibault-Northfield, MN Cloud, MN lmar, MN nkato-North Mankato, MN xandria, MN nidii. MN chinson, MN atonna, MN neapolis-St. Paulomington, MN-WI not. ND narck, ND liston, ND kinson, ND soula, MT ispell, MT sacola-Ferry Pass-Brent, FL bile, AL stview-Fort Walton Beach-

Destin, FL Daphne-Fairhope-Foley, AL

Monroe, LA El Dorado, AR Bastrop, LA Natchez, MS-LA Ruston, LA Salinas, CA San Jose-Sunnyvale-Santa Clara, CA Santa Cruz-Watsonville, CA Montgomery, AL Troy, AL Talladega-Sylacauga, AL Selma, AL Tullahoma-Manchester, TN

A E (Continued) Table

DMA CBSA Nashville TN Nashville-TN Nashville TN Lewisburg, TN Nashville TN Nashville TN Nashville TN Nashville TN Shelbyville, TN Nashville TN Cookeville, TN Nashville TN Paris, TN Nashville TN McMinnville, TN Nashville TN New Orleans LA New Orleans LA Picayune, MS New Orleans LA New Orleans LA New Orleans LA Bogalusa, LA New Orleans LA Hammond, LA New York NY New York NY NY-NJ-PA New York NY NY-NJ-PA New York NY Kingston, NY New York NY NY-NJ-PA New York NY NJ Norfolk-Portsmouth-Newport News VA News, VA-NC Norfolk-Portsmouth-Newport News VA Norfolk-Portsmouth-Newport News VA News, VA-NC Norfolk-Portsmouth-Newport News VA Norfolk-Portsmouth-Newport News VA Norfolk-Portsmouth-Newport News VA Norfolk-Portsmouth-Newport News VA Radford, VA North Platte NE North Platte, NE Odessa-Midland TX Odessa, TX Midland, TX Odessa-Midland TX Odessa-Midland TX Pecos, TX Odessa-Midland TX Big Spring, TX Odessa-Midland TX Andrews, TX Oklahoma City OK Oklahoma City OK Ponca City, OK Oklahoma City OK Enid, OK Oklahoma City OK Shawnee, OK Oklahoma City OK Woodward, OK Oklahoma City OK Elk City, OK Oklahoma City OK Stillwater, OK Oklahoma City OK Weatherford, OK Omaha NE Omaha NE Omaha NE Fremont, NE Columbus, NE Omaha NE Orlando-Daytona Beach-Melbourne FL Beach, FL Orlando-Daytona Beach-Melbourne FL Orlando-Daytona Beach-Melbourne FL The Villages, FL Orlando-Daytona Beach-Melbourne FL Ocala, FL Orlando-Daytona Beach-Melbourne FL FL. Ottumwa IA-Kirksville MO Ottumwa, IA Ottumwa IA-Kirksville MO Fairfield, IA Kirksville. MO Ottumwa IA-Kirksville MO Paducah KY-Cape Girardeau MO-Kennett, MO Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO-Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO-Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO-Paducah, KY-IL Harrisburg-Mount Vernon IL

Davidson-Murfreesboro-Franklin Lawrenceburg, TN Clarksville, TN-KY Bowling Green, KY Clarksville, TN-KY New Orleans-Metairie, LA Houma-Thibodaux, LA Gulfport-Biloxi-Pascagoula, MS Bridgeport-Stamford-Norwalk, CT New York-Newark-Jersey City, New York-Newark-Jersey City, New York-Newark-Jersey City, Allentown-Bethlehem-Easton, PA-Virginia Beach-Norfolk-Newport Elizabeth City, NC Virginia Beach-Norfolk-Newport Big Stone Gap, VA Kill Devil Hills, NC Roanoke Rapids, NC Blacksburg-Christiansburg-Oklahoma City, OK Omaha-Council Bluffs, NE-IA Omaha-Council Bluffs, NE-IA Deltona-Daytona Beach-Ormond Orlando-Kissimmee-Sanford, FL Palm Bay-Melbourne-Titusville, Carbondale-Marion, IL Mount Vernon, IL

G	able A.5 (Continued)
	DMA
	Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL
	Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL
	Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO-
	Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO-
	Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL
	Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL
	Paducah KY-Cape Girardeau MO- Harrisburg-Mount Vernon IL Paducah KY-Cape Girardeau MO-
	Harrisburg-Mount Vernon IL Panama City FL Panama City FL
	Panama City FL Parkersburg WV
	Parkersburg WV Peoria-Bloomington IL Peoria-Bloomington IL
	Peoria-Bloomington IL Peoria-Bloomington IL
	Peoria-Bloomington IL Philadelphia PA Philadelphia PA
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	Philadelphia PA
	Philadelphia PA Philadelphia PA
	Philadelphia PA Philadelphia PA Phoenix AZ
	Phoenix AZ Phoenix AZ
	Phoenix AZ Phoenix AZ Phoenix AZ
	Phoenix AZ Pittsburgh PA
	Pittsburgh PA Pittsburgh PA Pittsburgh PA
	Pittsburgh PA Portland OR
	Portland OR Portland OR
	Portland OR Portland OR
	Portland OR Portland OR Portland OR
	Portland OR Portland OR
	Portland OR Portland OR Portland OR
	Portland-Auburn ME Portland-Auburn ME
	Portland-Auburn ME Portland-Auburn ME

CBSA Poplar Bluff, MO Union City, TN-KY Mayfield, KY Paducah, KY-IL Sikeston, MO Cape Girardeau, MO-IL Martin, TN Cape Girardeau, MO-IL Union City, TN-KY Murray, KY Panama City, FL Crestview-Fort Walton Beach-Destin, FL Parkersburg-Vienna, WV Marietta, OH Peoria, IL Ottawa-Peru, IL Pontiac, IL Bloomington, IL Canton, IL Atlantic City-Hammonton, NJ Philadelphia-Camden-Wilmington, PA-NJ-DE-MD Allentown-Bethlehem-Easton, PA-NJ Dover, DE Reading, PA Philadelphia-Camden-Wilmington, PA-NJ-DE-MD Vineland-Bridgeton, NJ Philadelphia-Camden-Wilmington, PA-NJ-DE-MD Ocean City, NJ Trenton, NJ Payson, AZ Phoenix-Mesa-Scottsdale, AZ Safford, AZ Prescott, AZ Lake Havasu City-Kingman, AZ Show Low, AZ Flagstaff, AZ Pittsburgh, PA Indiana, PA Morgantown, WV Oil City, PA New Castle, PA Prineville, OR Hood River, OR Portland-Vancouver-Hillsboro, OR-WA Portland-Vancouver-Hillsboro, OR-WA Longview, WA Salem, OR Albany, OR The Dalles, OR Newport, OR La Grande, OR Hermiston-Pendleton, OR Astoria, OR Portland-South Portland, ME Augusta-Waterville, ME Berlin, NH-VT Lewiston-Auburn, ME

DMA

Providence RI-New Bedford MA Providence RI-New Bedford MA Quincy IL-Hannibal MO-Keokuk IA Ouincy IL-Hannibal MO-Keokuk IA Raleigh-Durham (Fayetteville) NC Rapid City SD Rapid City SD Rapid City SD Rapid City SD Reno NV Richmond-Petersburg VA Richmond-Petersburg VA Richmond-Petersburg VA Roanoke-Lynchburg VA Roanoke-Lynchburg VA Roanoke-Lynchburg VA Roanoke-Lynchburg VA Roanoke-Lynchburg VA Roanoke-Lynchburg VA Rochester MN-Mason City IA-Austin MN Rochester MN-Mason City IA-Austin MN

Rochester MN-Mason City IA-Austin MN Rochester MN-Mason City IA-Austin MN Rochester NY Rockford IL Rockford IL Rockford IL Rockford IL Sacramento-Stockton-Modesto CA Sacramento-Stockton-Modesto CA

Sacramento-Stockton-Modesto CA Sacramento-Stockton-Modesto CA Sacramento-Stockton-Modesto CA Sacramento-Stockton-Modesto CA Sacramento-Stockton-Modesto CA Salisbury MD Salisbury MD Salisbury MD Salt Lake City UT Salt Lake City UT

CBSA

Quincy, IL-MO

Hannibal, MO

Jacksonville, IL

Quincy, IL-MO

Roanoke Rapids, NC

Macomb, IL

Wilson, NC

Sanford, NC

Dunn, NC Goldsboro, NC

Raleigh, NC

Oxford, NC Rapid City, SD

Spearfish, SD

Sheridan, WY

Susanville, CA

Reno, NV

Fallon, NV

Fernley, NV

Carson City, NV

Winnemucca, NV

Charlottesville, VA

Richmond, VA

News, VA-NC

Roanoke, VA

Radford, VA

Danville, VA

Charlottesville, VA

Albert Lea, MN

Mason City, IA

Rochester, MN

Rochester, NY

Rockford, IL

Freeport, IL

Rochelle, IL

Yuba City, CA

Stockton-Lodi, CA

Salisbury, MD-DE

Salisbury, MD-DE Ogden-Clearfield, UT

Salt Lake City, UT

Heber, UT

Evanston, WY

Logan, UT-ID

Summit Park, UT

St. George, UT

Vernal, UT

Elko, NV

Price, UT

Cambridge, MD

Vallejo-Fairfield, CA

Arcade, CA

Sonora, CA

Modesto, CA Truckee-Grass Valley, CA

Sacramento-Roseville-Arden-

Dixon, IL

Austin, MN

Lynchburg, VA

Martinsville, VA

Scottsbluff, NE

Henderson, NC

Fayetteville, NC

Rocky Mount, NC

Durham-Chapel Hill, NC

Gardnerville Ranchos, NV

Virginia Beach-Norfolk-Newport

Blacksburg-Christiansburg-

Providence-Warwick, RI-MA

Providence-Warwick, RI-MA

Fort Madison-Keokuk, IA-IL-MO

Fort Madison-Keokuk, IA-IL-MO

Fort Madison-Keokuk, IA-IL-MO

Pinehurst-Southern Pines, NC

DMA	CBSA
Salt Lake City UT	Logan, UT-ID
Salt Lake City UT	Cedar City, UT
Salt Lake City UT	Rock Springs, WY
Salt Lake City UT	Provo-Orem, UT
San Angelo TX	San Angelo, TX
San Antonio TX	San Antonio-New Braunfels, TX
San Antonio TX	Victoria, TX
San Antonio TX	Uvalde, TX
San Antonio TX	Eagle Pass, TX
San Antonio TX	Kerrville, TX
San Antonio TX	Del Rio, TX
San Diego CA	San Diego-Carlsbad, CA
San Francisco-Oakland-San Jose CA	San Francisco-Oakland-Hayward
San Hancisco-Oakland-San Sose CA	CA
San Francisco-Oakland-San Jose CA	San Jose-Sunnyvale-Santa Clara, CA
San Francisco-Oakland-San Jose CA	Ukiah, CA
San Francisco-Oakland-San Jose CA	Santa Rosa, CA
San Francisco-Oakland-San Jose CA	Napa, CA
San Francisco-Oakland-San Jose CA	Clearlake, CA
Santa Barbara-Santa Maria-San Luis	Santa Maria-Santa Barbara, CA
Obispo CA	
Santa Barbara-Santa Maria-San Luis	San Luis Obispo-Paso Robles-
Obispo CA	Arroyo Grande, CA
Savannah GA	Statesboro, GA
Savannah GA	Jesup, GA
Savannah GA	Brunswick, GA
Savannah GA	Hinesville, GA
Savannah GA	Hilton Head Island-Bluffton-
Constant CA	Beaufort, SC
Savannah GA	Vidalia, GA
Savannah GA	Savannah, GA
Seattle-Tacoma WA	Aberdeen, WA
Seattle-Tacoma WA	Wenatchee, WA
Seattle-Tacoma WA	Seattle-Tacoma-Bellevue, WA
Seattle-Tacoma WA	Bellingham, WA
Seattle-Tacoma WA	Oak Harbor, WA
Seattle-Tacoma WA	Mount Vernon-Anacortes, WA
Seattle-Tacoma WA	Shelton, WA
Seattle-Tacoma WA	Port Angeles, WA
Seattle-Tacoma WA	Bremerton-Silverdale, WA
Seattle-Tacoma WA	Centralia, WA
Seattle-Tacoma WA	Olympia-Tumwater, WA
Sherman TX-Ada OK	Sherman-Denison, TX
Sherman TX-Ada OK	Durant, OK
Sherman TX-Ada OK	Ada, OK
Sherman TX-Ada OK	Ardmore, OK
Shreveport LA	Shreveport-Bossier City, LA
Shreveport LA	Mount Pleasant, TX
Shreveport LA	Texarkana, TX-AR
Shreveport LA	Natchitoches, LA
Shreveport LA	Texarkana, TX-AR
Shreveport LA	Magnolia, AR
Shreveport LA	Magnona, AK Marshall, TX
Sioux City IA	Sioux City, IA-NE-SD
	Sioux City, IA-NE-SD Sioux City, IA-NE-SD
Sioux City IA Sioux City IA	
Sioux City IA	Spirit Lake, IA
Sioux City IA	Sioux City, IA-NE-SD
Sioux City IA	Spencer, IA
Sioux City IA	Norfolk, NE
Sioux City IA	Storm Lake, IA
Sioux Falls(Mitchell) SD	Sioux Falls, SD
Sioux Falls(Mitchell) SD	Pierre, SD
Sioux Falls(Mitchell) SD	Yankton, SD
Sioux Falls(Mitchell) SD	Mitchell, SD
Sioux Falls(Mitchell) SD	Brookings, SD
Sioux Falls(Mitchell) SD	Huron, SD
Sioux Falls(Mitchell) SD	Aberdeen, SD
Sioux Falls(Mitchell) SD	Vermillion, SD
Sioux Falls(Mitchell) SD	Watertown, SD
Sioux Falls(Mitchell) SD	Worthington, MN
South Bend-Elkhart IN	Plymouth, IN
South Bend-Elkhart IN	Elkhart-Goshen, IN
South Bend-Flkhart IN	Warsaw IN

Warsaw, IN

South Bend-Elkhart IN

Table A.5 (Continued)

DMA

South Bend-Elkhart IN South Bend-Elkhart IN South Bend-Elkhart IN Spokane WA Springfield MO Springfield-Holyoke MA Springfield-Holyoke MA St. Joseph MO St. Joseph MO St. Joseph MO St. Louis MO St. Louis MO St. Louis MO St. Louis MO Syracuse NY Syracuse NY Syracuse NY Svracuse NY Syracuse NY Tallahassee FL-Thomasville GA Tallahassee FL-Thomasville GA Tallahassee FL-Thomasville GA Tallahassee FL-Thomasville GA Tampa-St. Petersburg (Sarasota) FL Tampa-St. Petersburg (Sarasota) FL Tampa-St. Petersburg (Sarasota) FL

Tampa-St. Petersburg (Sarasota) FL

Tampa-St. Petersburg (Sarasota) FL Tampa-St. Petersburg (Sarasota) FL Terre Haute IN Terre Haute IN Terre Haute IN Toledo OH Topeka KS Topeka KS Topeka KS Topeka KS Traverse City-Cadillac MI Tri-Cities TN-VA Tri-Cities TN-VA Tri-Cities TN-VA Tri-Cities TN-VA Tri-Cities TN-VA Tri-Cities TN-VA Tucson (Sierra Vista) AZ Tucson (Sierra Vista) AZ

CBSA South Bend-Mishawaka, IN-MI South Bend-Mishawaka, IN-MI Niles-Benton Harbor, MI Walla Walla, WA Pullman, WA Sandpoint, ID Lewiston, ID-WA Spokane-Spokane Valley, WA Lewiston, ID-WA Coeur dAlene, ID Moscow, ID Moses Lake, WA Othello, WA Springfield, MO Harrison, AR Mountain Home, AR Branson, MO Fort Leonard Wood, MO West Plains, MO Rolla, MO Lebanon, MO Springfield, MA Greenfield Town, MA St. Joseph, MO-KS Maryville, MO St. Joseph, MO-KS St. Louis, MO-IL St. Louis, MO-IL Farmington, MO Centralia, IL Seneca Falls, NY Cortland, NY Syracuse, NY Ithaca, NY Auburn, NY Tallahassee, FL Valdosta, GA Thomasville, GA Bainbridge, GA Sebring, FL Homosassa Springs, FL Tampa-St. Petersburg-Clearwater, FL North Port-Sarasota-Bradenton, FL Lakeland-Winter Haven, FL Wauchula, FL Terre Haute, IN Vincennes, IN Washington, IN Defiance, OH Findlay, OH Toledo, OH Fremont, OH Adrian, MI Tiffin, OH Port Clinton, OH Manhattan, KS Topeka, KS Emporia, KS Junction City, KS Cadillac, MI Sault Ste. Marie, MI Traverse City, MI Big Rapids, MI Ludington, MI Johnson City, TN Big Stone Gap, VA Kingsport-Bristol-Bristol, TN-VA Richmond, VA Kingsport-Bristol-Bristol, TN-VA Greeneville, TN Sierra Vista-Douglas, AZ Nogales, AZ

DMA	CBSA
Tucson (Sierra Vista) AZ	Tucson, AZ
Tulsa OK	Bartlesville, OK
Tulsa OK	Tulsa, OK
Tulsa OK	Coffeyville, KS
Tulsa OK	Tahlequah, OK
Tulsa OK	Muskogee, OK
	0 11
Tulsa OK	McAlester, OK
Twin Falls ID	Hailey, ID
Twin Falls ID	Burley, ID
Twin Falls ID	Twin Falls, ID
Tyler-Longview(Lufkin & Nacogdoches) TX	Jacksonville, TX
Tyler-Longview(Lufkin & Nacogdoches) TX	Huntsville, TX
Tyler-Longview(Lufkin & Nacogdoches) TX	Tyler, TX
Tyler-Longview(Lufkin & Nacogdoches) TX	Nacogdoches, TX
Tyler-Longview(Lufkin & Nacogdoches) TX	Longview, TX
Tyler-Longview(Lufkin & Nacogdoches) TX	Lufkin, TX
Utica NY	Oneonta, NY
Utica NY	Utica-Rome, NY
Victoria TX	Victoria, TX
Waco-Temple-Bryan TX	Killeen-Temple, TX
Waco-Temple-Bryan TX	College Station-Bryan, TX
Waco-Temple-Bryan TX	Waco, TX
Washington DC (Hagerstown MD)	Winchester, VA-WV
Washington DC (Hagerstown MD)	Washington-Arlington-
	Alexandria, DC-VA-MD-WV
Washington DC (Hagerstown MD)	Washington-Arlington-
	Alexandria, DC-VA-MD-WV
Washington DC (Hagerstown MD)	Washington-Arlington-
	Alexandria, DC-VA-MD-WV
Washington DC (Hagerstown MD)	Cumberland, MD-WV
Washington DC (Hagerstown MD)	Winchester, VA-WV
Washington DC (Hagerstown MD)	Hagerstown-Martinsburg, MD-
	WV
Washington DC (Hagerstown MD)	Washington-Arlington-
	Alexandria, DC-VA-MD-WV
Washington DC (Hagerstown MD)	Cumberland, MD-WV
Washington DC (Hagerstown MD)	Chambersburg-Waynesboro, PA
Washington DC (Hagerstown MD)	California-Lexington Park, MD
Washington DC (Hagerstown MD)	Hagerstown-Martinsburg, MD-
	WV
Watertown NY	Ogdensburg-Massena, NY
Watertown NY	Watertown-Fort Drum, NY
Wausau-Rhinelander WI	Wausau, WI
Wausau-Rhinelander WI	Merrill, WI
Wausau-Rhinelander WI	Wisconsin Rapids-Marshfield, W
Wausau-Rhinelander WI	Stevens Point, WI
West Palm Beach-Ft. Pierce FL	Miami-Fort Lauderdale-West
	Palm Beach, FL
West Palm Beach-Ft. Pierce FL	Port St. Lucie, FL
West Palm Beach-Ft. Pierce FL	Sebastian-Vero Beach, FL
West Palm Beach-Ft. Pierce FL	Okeechobee, FL
Wheeling WV-Steubenville OH	Weirton-Steubenville, WV-OH
Wheeling WV-Steubenville OH	,
0	Weirton-Steubenville, WV-OH
Wheeling WV-Steubenville OH	Wheeling, WV-OH
Wheeling WV-Steubenville OH	Wheeling, WV-OH
Wichita Falls TX & Lawton OK	Wichita Falls, TX
Wichita Falls TX & Lawton OK	Lawton, OK
Wichita Falls TX & Lawton OK	Altus, OK
Wichita Falls TX & Lawton OK	Vernon, TX
Wichita Falls TX & Lawton OK	Duncan, OK
Wichita-Hutchinson KS	Wichita, KS
Wichita-Hutchinson KS	Liberal, KS
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Wichita-Hutchinson KS	Salina, KS
Wichita-Hutchinson KS	McPherson, KS
Wichita-Hutchinson KS	Dodge City, KS
Wichita-Hutchinson KS	Arkansas City-Winfield, KS
Wichita-Hutchinson KS	Great Bend, KS
Wichita-Hutchinson KS	Garden City, KS
Wichita-Hutchinson KS	Hutchinson, KS
Wichita-Hutchinson KS	Hays, KS
	-
Wilkes Barre-Scranton PA	Selinsgrove, PA
Wilkes Barre-Scranton PA	Scranton–Wilkes-Barre–Hazletor
	PA

Sunbury, PA

Wilkes Barre-Scranton PA

Table A.5 (Continued)

DMA Wilkes Barre-Scranton PA Wilkes Barre-Scranton PA

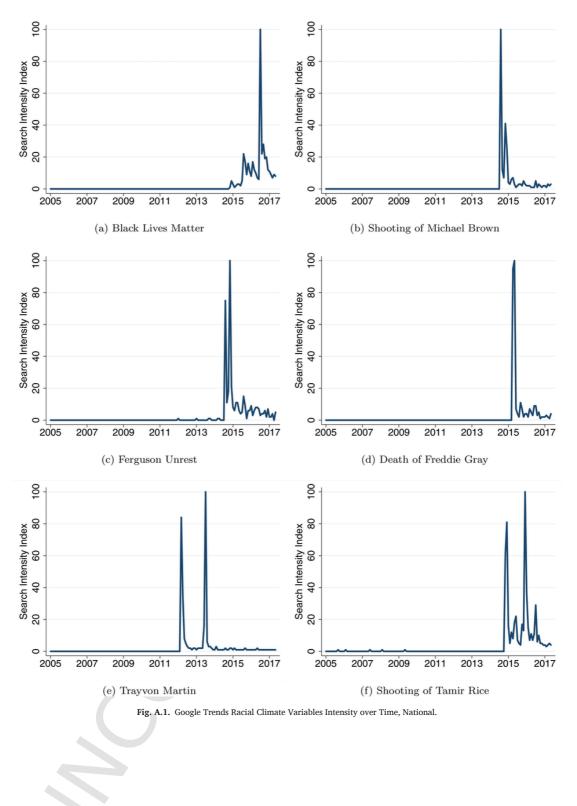
Yakima-Pasco-Richland-Kennewick WA Yakima-Pasco-Richland-Kennewick WA Yakima-Pasco-Richland-Kennewick WA Yakima-Pasco-Richland-Kennewick WA Yakima-Pasco-Richland-Kennewick WA

Youngstown OH Youngstown OH

Yuma AZ-El Centro CA Yuma AZ-El Centro CA Zanesville OH Bloomsburg-Berwick, PA East Stroudsburg, PA Lewisburg, PA Pottsville, PA Williamsport, PA Lock Haven, PA Sayre, PA Allentown-Bethlehem-Easton, PA-NJ Wilmington, NC Myrtle Beach-Conway-North Myrtle Beach, SC-NC Yakima, WA Hermiston-Pendleton, OR Ellensburg, WA Walla Walla, WA Kennewick-Richland, WA Youngstown-Warren-Boardman, OH-PA Youngstown-Warren-Boardman, OH-PA Salem, OH Yuma, AZ El Centro, CA Zanesville, OH

CBSA

We use Sood (2016) to assign DMA information to counties and use a crosswalk from Missouri Census Data Center (2012) to translate county level information into CBSAs. A downloadable version of our constructed crosswalk can be accessed at http://www.yelowitz.com/racialclimate.



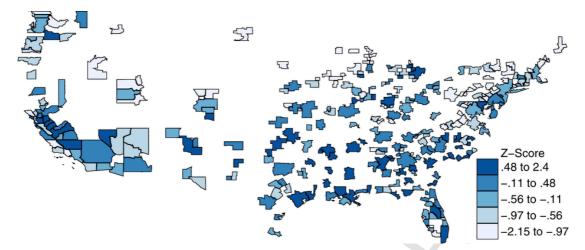


Fig. A.2. Crime Intensity Z-score by Metro Area 2012 Note: Crime statistics originate from the FBI Unified Crimes Report.

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