From Anti-China Rhetoric to Anti-Asian Behavior: The Social and Economic Cost of “Kung Flu”

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Abstract

Discrimination and violence directed towards Asian Americans in the United States increased dramatically following the onset of the COVID-19 pandemic. In this paper, we examine consumer discrimination against businesses associated with Asian Americans. Leveraging the pandemic as an exogenous shock to Americans’ level of anti-Chinese sentiment, we utilize a series of analyses combining survey data, online search trends, and consumer cellular device mobility data to measure the effects of this shock on consumer discrimination against Chinese and other Asian restaurants. Survey and search data show that attitudes towards Chinese and non-Chinese Asian food declined precipitously during the pandemic, and this change in attitudes was driven by a mix of assigning blame for COVID-19 spread to Asians and experiencing fear of Chinese food. Analysis of cellular phone mobility data shows Asian restaurants suffered a 18.4% drop (95% C.I.: -15.9% to -20.8%) in traffic relative to non-Asian restaurants in the pandemic period. We explore heterogeneity in these effects by political affiliation and find strong correlation between support of former President Trump and avoidance of Asian restaurants. The results are consistent with the role of out-group homogeneity and ethnic misidentification as drivers of spillover effects of anti-Chinese sentiment on non-Chinese Asian restaurant traffic. This work documents some of the unique economic challenges faced by Asian Americans during the COVID-19 pandemic and has substantial implications for the study of consumer discrimination and stigmatization in public health communications.
1 Introduction

Scholars have long documented patterns of labor discrimination against people of color in the U.S. (Bertrand and Mullainathan 2004; Pager 2003; Kuppuswamy and Younkin 2020; Acquisti and Fong 2020; Chavez 2021; Broyles and Fenner 2010). These forms of discrimination have substantial negative impacts on communities of color, denying social mobility to historically marginalized groups (Dustmann and Theodoropoulos 2010; Ghazaryan 2019). Even as some forms of discrimination have decreased over the recent decades (Quillian, Lee and Honoré 2020), racial discrimination in labor markets remains stubbornly persistent (Mason et al. 2005; Quillian et al. 2017). In this paper, we examine a closely related but much less studied economic phenomenon – consumer discrimination against businesses, especially small businesses associated with Asian Americans during the COVID-19 pandemic.

Historically, marginalized groups have responded to the challenges of labor discrimination by starting their own businesses providing unique goods and services. Many Asian immigrants to the United States achieved economic success using this model by starting businesses such as laundromats and ethnic restaurants (Liu 2015). In the late 19th century, the U.S. heavily encouraged Asian immigration to supply cheap labor for manufacturing their burgeoning railroad system (White 2011). These immigrants founded America’s first Chinese restaurants. By the early 20th century, both hostility towards Chinese immigrants and Americans’ appetite for Americanized Chinese food had increased substantially (Liu 2015). The 1970’s saw a further resurgence of American interest in "authentic” Chinese food (Liu 2015). While restaurant entrepreneurship remains a pathway to economic success for Asian Americans, discrimination by consumers can negatively impact the efficacy of this pathway, especially during periods of heightened anti-Asian sentiment.

Marginalization, stigmatization, and even violence towards minority groups has been a frequent historical response to public health (McCauley, Minsky and Viswanath 2013), terrorism (Chandrasekhar 2003), and economic (Johnson 1996) events. While public blame (i.e., rhetoric
from politicians and mainstream media) is typically directed towards a singular foreign group, domestic hate crimes in response to these events often scapegoat American minorities and related groups that are mistakenly targeted. Tragically, Balbir Singh Sodhi, a Sikh American, was murdered after being mistaken for Arab Muslim in the aftermath of the 9/11 attacks, and Vincent Chin, a Chinese American, was murdered after being mistaken for Japanese amidst rising blue collar unemployment and American-Japanese economic competition in the early 1980s. Research on out-group homogeneity bias (Quattrone and Jones 1980) suggested that people tend to view out-group members as more similar to one another, as opposed to in-group members. Adding to this bias, scholars have identified a cross-race effect, which is the tendency to recognize faces that belong to one’s own racial group while having difficulty differentiating members of the cross-racial group (Chance and Goldstein 1981; Lebrecht et al. 2009). Further, individuals with less cross-race exposure are more likely to harbor prejudicial attitudes towards other groups (Bursztyn et al. 2021; Zebrowitz, White and Wienek 2008). Taken together, these effects can allow prejudicial attitudes towards an out-group (Asian Americans, in our case) to drive prejudicial actions directed towards domestic minority groups.

In this paper, we examine the role of consumer discrimination against Asian businesses in the United States in the context of the COVID-19 pandemic. Previous studies have found that consumer discrimination substantially decreases small business ownership and self-employment among people of color in the U.S. (Borjas and Bronars 1989; Fairlie 1999; Holzer and Ilanfeldt 1998; Liu, Liao and Ming 2020), as does discrimination by funding sources (Younkin and Kuppuswamy 2018). Among small businesses, consumer discrimination is especially salient for restaurants which specialize in ethnic cuisines, as the food’s ethnic origin is a key aspect of product differentiation. We leverage the COVID-19 pandemic as an exogenous shock to rhetoric and attitudes about Asian Americans in the U.S. Soon after COVID-19 began spreading in the U.S., then-President Trump and the Republican Party began a concerted effort to blame China for the virus. Political and media figures used the phrases "China virus" and "Kung flu" to describe the coronavirus, in contradiction to World Health Organization (WHO)
guidance (World Health Organization, 2015) and prior research demonstrating how disease naming can lead to stigmatization, particularly of marginalized and minority groups (Perry and Donini-Lenhoff, 2010; McCauley, Minsky and Viswanath, 2013; Keusch, Wilentz and Kleinman, 2006). The effects of this rhetoric did not go unnoticed – the number of anti-Asian hate crimes spiked during the pandemic (Gover, Harper and Langton, 2020; Tessler, Choi and Kao, 2020), impacting both Chinese Americans and the broader Asian American community. Self-reports and surveys from Asian American citizens and business owners have suggested that the community has seen disproportionate, negative impacts during the pandemic, though quantification is challenging due to the localized nature of these studies and potential for selection biases.

Using web search, survey, and movement data, we document and quantify the impact of anti-China sentiment on both Chinese Americans and spillover effects on other Asian American populations. First, we use web search data to demonstrate the exogenous shock of the COVID-19 pandemic on American conceptions of China and Chinese food. Next, movement data allows us to measure the effect of anti-Asian attitudinal shifts on consumer discrimination against both Chinese and other Asian restaurants, quantifying the economic effect of anti-Chinese rhetoric on Asian American communities. Finally, we use survey data to explore different underlying mechanisms that influence these shifts.

We find clear evidence of sharp attitudinal shifts against China, Chinese Americans, and Asian Americans as a whole during the COVID-19 pandemic. These shifts have real economic consequences for Asian American communities, as people who have internalized this rhetoric have heightened concerns over the safety of Chinese products, most prominently Chinese restaurants. Furthermore, many Americans are not good at differentiating between Chi-

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1 Asian-owned businesses say they’re reeling from hate and violence, operating in fear. Last accessed Feb 15, 2022. [https://www.pbs.org/newshour/economy/asian-owned-businesses-say-theyre-reeling-from-hate-and-violence-operating-in-fear](https://www.pbs.org/newshour/economy/asian-owned-businesses-say-theyre-reeling-from-hate-and-violence-operating-in-fear)

2 Stop AAPI Hate National Report. Last accessed Feb 15, 2022. [https://stopaapihate.org/national-report-through-september-2021/](https://stopaapihate.org/national-report-through-september-2021/)

3 Asian American Businesses: Identifying Gaps and Supporting Recovery. Last accessed Feb 15, 2022. [http://www.aasc.ucla.edu/resources/policyreports/ABA_Survey_Brief_Recovery21.pdf](http://www.aasc.ucla.edu/resources/policyreports/ABA_Survey_Brief_Recovery21.pdf)
nese Americans and other East Asian Americans, putting the larger Asian American community at risk from anti-Chinese sentiment and behavior. It is not surprising that we find an early and disproportionate drop in business suffered by Chinese restaurants during the COVID-19 crisis, but also for other non-Chinese Asian American restaurants, suggesting that the increase in anti-China sentiment incurred not only a social cost for Asian Americans at large, but it also incurred a real economic cost.

Our contributions are three-fold. First, we examine the role of racially-salient events in shaping levels of consumer discrimination against businesses. The accessibility of race as a dominant category for in-group and out-group distinction has been shown to influence various outcomes for the minoritized groups, ranging from housing discrimination for Arab-Americans (Gaddis and Ghoshal 2015) to evaluative bias and discrimination against African American job applicants and entrepreneurs (Mobasseri 2019, Gorbatai, Younkin and Burtch 2021). Our work establishes the pandemic-driven blaming of Asians as one instantiation of mega-threats, which refers to negative, identity-relevant societal events that receive significant media attention (Leigh and Melwani 2019) in the context of entrepreneurship and consumers. Entrepreneurship has historically served as a viable exit strategy for populations facing labor discrimination. However, for businesses serving customers outside its ethnic community of origin, consumer discrimination can hamper the businesses’ economic viability. If levels of consumer discrimination fluctuate based on current events, otherwise profitable businesses may quickly become unviable during periods of high discrimination and violence. Furthermore, unpredictability of demand for the businesses’ products may also serve as a barrier to entry for potential entrepreneurs.

While our results focus on consumer discrimination against Asian restaurants in the wake of the pandemic, these findings have relevance in a much broader context. Restaurants are a bellwether for broader anti-minority sentiment and its impact on small businesses due to their ubiquity and easy association with an ethnic group. Other small businesses such as dental of-
fices, lawyers, doctors, accountants, are also easily affiliated with an ethnic group due to typical naming conventions, and our conclusions on how restaurant avoidance is not explained by consumer health safety concerns have substantial implications for consumer discrimination against these businesses as well. There are numerous historical examples of anti-ethnic sentiment due to major geopolitical events, including anti-Arab/Islam and anti-French (“Freedom Fries”) sentiment in the aftermath of the 9/11 attacks and anti-Japanese sentiment amidst rising American-Japanese competition in the 1980s. However, despite numerous reports on attacks on minorities and anecdotal reports of consumer discrimination, there are significant difficulties with quantifying the impacts these crises had on minority-owned businesses. The recent pandemic has coincided with major technological advances in location data availability, making this the first time that researchers can conduct granular, quantitative analysis of small business impacts and misidentification-related spillovers which were impossible during previous crises.

Second, our research uncovers the consumer discrimination that may have gone unnoticed otherwise. Recent research has documented the rise of racial discrimination and prejudice against Asians and Asian Americans, which resulted in poorer mental and physical health (Lee and Waters [2020]) and marginalization (Li and Nicholson Jr. [2021]) in these groups. Other studies on anti-Asian discrimination during the pandemic has relied heavily on data from self-reported incidents of hate against Asian Americans (Horse et al. [2021]) and police statistics on violent attacks, while in contrast our study examines more subtle social and economic impacts of discrimination which are highly relevant to the lives of Asian Americans yet might not rise directly to the level of hate or violence. This study avoids the selection biases of self- or police reporting where immigrant groups may be hesitant to contact authorities (Yun and Mueller [2011]), and it utilizes a big data approach to the question of discrimination covering hundreds of billions of trips to six million locations throughout the US in comparison to the approximately 9,000 self-reported hate incidents and only 279 hate crimes identified by the

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4 Anti-Asian hate crimes rose 73% last year, updated FBI data says. Last accessed Feb 15, 2022. https://www.nbcnews.com/news/asian-america/anti-asian-hate-crimes-rose-73-last-year-updated-fbi-data-says-rcna3741
Third and finally, our study contributes to the psychological theories of collective blame by highlighting the role of ethnic misclassification in consumer discrimination. By examining the costs incurred on both Chinese and non-Chinese Asian restaurants, we capture the process of blame spillover from businesses associated with one ethnic group to the larger tangentially-related populations. This spillover effect, at least partly fueled by the “yellow peril” stereotype for Asians in general (Tessler, Choi and Kao 2020), lends support to the theory of collective blame (Denson et al. 2006; Lickel et al. 2006; Stenstrom et al. 2008) that arises from out-group homogeneity bias and cross-race effects. To the best of our knowledge, our study is first in examining these spillover effects via ethnic misclassification in the context of consumer discrimination in an ecologically valid setting. For instance, cross-race effects have been found in people’s perception of out-group members’ faces. Our research extends this line of work by showing that a similar pattern occurs with consumer perception of out-group-owned businesses as well. Studying consumer discrimination based in ethnic and racial misclassification also advances our understanding the experiences of individuals and entities that are regularly misclassified. For instance, research on racial misclassification (Campbell and Troyer 2007) found the negative psychological consequences for American Indians who were often misclassified for a different racial group. Our research stresses the prominence of a similar mechanism for Asians who are not ethnically Chinese, and broadens it to encompass social and economic costs. Understanding the role of ethnic misclassification in discrimination allows us to better understand the cognitive processes behind discriminatory behavior.

The rest of the paper is organized as follows. Section 2 focuses on background and theoretical perspectives that inform our hypotheses for study. Section 3 describes our data sources. Section 4 describes our methodology and results. We confirm the role of the COVID-19 pandemic as an exogenous shock to anti-Asian sentiment and highlight the unique effect of this shock on Chinese and other Asian restaurants. We identify fear of Chinese food safety, blame
of Asian Americans for COVID-19, and ethnic misclassification of restaurants as possible psychological mechanisms for this effect. The fifth and final section describes our conclusions and suggests further directions for this research.

2 Background

Innuendos of "uncleanness" and disease have long been used to stigmatize marginalized populations and justify their expulsion. From blaming European Jews for the Bubonic Plague in the Middle Ages (Jedwab, Johnson and Koyama 2019) to claiming that immigrants crossing the United States’ southern border are 'diseased' majority groups have long misattributed blame for diseases to "undesirable" minority populations. In the U.S., portrayal of immigrants as dangerous and diseased shaped immigration policy for decades (Kraut 1995; Markel and Stern 1999). Modern-day attitudes towards disease threat are closely related to anti-immigration attitudes (Green et al. 2010; Krings et al. 2012). This carries forward into the present with rhetoric falsely trying to tie the fourth wave of COVID-19 illness to undocumented immigrants coming through the U.S.’ southern border.

These stigmas have been especially prevalent in the treatment of Asian Americans in the 19th and 20th centuries. Medical examinations of Chinese immigrants arriving at Angel Island were frequently harsher than those of Europeans at Ellis Island, and the rejection rate for Chinese immigrants was at least five times higher than for immigrants at Ellis island (Markel and Stern 1999). These attitudes extended towards Asian cultural products in the U.S., especially Chinese food. While early Chinese restaurants founded in San Francisco during the California Gold Rush were praised as tasty and economical, the rising tide of Sinophobia soon led to rumors of the consumption of rats and snakes as Chinese delicacies, leading to a decline in Chi-

5 Anti-Immigrant Movement Has History of Politicizing Disease. Last accessed May 15, 2022. https://www.splcenter.org/hatewatch/2020/04/21/anti-immigrant-movement-has-history-politicizing-disease

6 https://www.msn.com/en-us/news/politics/politifact-declares-desantis-e2-80-99-claims-covid-rise-is-due-to-illegal-border-crossings-is-e2-80-98false-e2-80-99/ar-AAN1TqS?ocid=uxbndlbng
nese restaurant popularity (Liu 2015). More recently, the popularization of the term "Chinese Restaurant Syndrome", based in medically-unfounded fears over the effects of Monosodium Glutamate (MSG) (Wahlstedt et al. 2021; Mosby 2009), is another example of cultural discrimination against Chinese restaurants.

Portrayals of the COVID-19 crisis echoed many of these historical patterns of negative attitudes and discrimination against this group in the face of public health concerns. The COVID-19 virus was regularly referred to as the “Chinese virus” or “China virus”, both offline (Masters-Waage, Jha and Reb 2020) and online (Budhwani and Sun 2020; Hswen et al. 2021). Donald Trump and Trump-supporting Republicans and media began to refer to COVID-19 as the “China virus” and explicitly blame the pandemic on China despite consistent criticism from Asian Americans and left-leaning media outlets. Between March 16, 2020 and Jan 3, 2021, Trump used the phrase “China virus” or “Chinese virus” in 54 separate tweets. This explicit blame on China by the politicians was then reinforced and further spread into the mainstream media outlets that rebroadcast this rhetoric. Recent research has demonstrated how social media also plays a vital role in building and reproducing negative sentiment against marginalized groups. For example, recent studies have examined the spread of anti-Muslim sentiment through hashtags on Twitter (Miro-Llinares and Rodriguez-Sala 2016; Müller and Schwarz 2021). Similarly, two studies to date have demonstrated the link between hashtags that used the language of this rhetoric (#Chinesevirus, for instance) and anti-Asian sentiments on Twitter (Hswen et al. 2021; Pei and Mehta 2020). Although establishing a causal relationship between the political rhetoric and anti-Asian sentiment is beyond the scope of our current research, this suggests a possibility that anti-Asian sentiment initiated by the political figures may have spread broadly and quickly on both mainstream media and social media.

Against this backdrop, latest research adds to the growing evidence for the Asian Americans blamed for the COVID-19 pandemic. For instance, Reny and Barreto (2022) finds that among Americans, fear of COVID-19 is highly correlated with anti-Asian attitudes and sup-
port for anti-Asian immigration policies. Furthermore, an experimental study has shown that emphasizing the origin of COVID-19 to China increased anti-Asian sentiment and xenophobia (Dhanani and Franz 2021). Following this logic, we argue that the COVID-19 pandemic served as an exogenous shock\(^7\) that gave voice to politicians who were increasingly making explicitly racial appeals by singling out marginalized groups (Stephens-Dougan 2020), thereby contributing to the rise of anti-Asian attitudes within the U.S.

Drawing from the psychological theory of blame, we put forth the idea that not only was the country of origin directly blamed for the pandemic, but so were other groups often mislabeled as Chinese that subsequently became the targets of collective blame. Alicke’s (2000) model of culpable control suggests that blame is the outcome of psychological processes driven by cognitive and motivational biases, which may be independent of a party’s actual liability (Alicke 2000; Hart and Honore 1985; Prosser and Wade 1971). Further, his theory posits that people may assign blame based on the proximity criteria. To the extent that the associated party is viewed as sufficiently proximate to the party that is presumed to be guilty, it is possible that even the group (i.e., businesses owned by Chinese Americans) that is not blamed for the harm may still be cognitively implicated. While there were legitimate questions about whether or not food could carry COVID-19 in the early days of the pandemic, this concern alone would not be sufficient if some restaurants were disproportionately affected by the COVID-19 pandemic than others, as there was not a legitimate concern about any particular type of food. Facing COVID-19 as an exogenous shock and the anti-Asian rhetoric and attitudes that followed, we hypothesize that:

**H1: During the COVID-19 crisis, Chinese restaurants will see a substantial drop in visits relative to non-Asian restaurants.**

Much of the consumer research on blame spillover has been documented in the context of

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\(^7\) An exogenous shock refers to an unexpected or unpredictable event; for instance, exogenous events that come from outside of the economic system (such as wars, natural disasters, a global pandemic, and new technology) has an impact on economic or consumer behavior.
brands that were implicated in moral transgressions. This effect is found to be more prominent to the brands consumers identify as similar to the brand that was implicated in ethical scandal (Borah and Tellis 2016; Paruchuri and Misangyi 2015; Trump and Newman 2017). Based on this literature, we further argue that how consumers ascribe blame on the associated groups may contribute to the struggles of other Asian restaurants that are not Chinese. This blame spillover may occur via two channels. First, it is possible that non-Chinese entities were incorrectly identified as Chinese, thereby being erroneously blamed for the pandemic. Second, it is also plausible that non-Chinese entities were correctly identified (which we call “ethnic misidentification hypothesis” detailed below), but viewed as a proximate entity under the broader umbrella of Asian race and ethnicity, thereby being categorically avoided. Both possibilities support the idea that other Asian restaurants will be negatively affected at the early onset of COVID-19 in the U.S., leading to our second Hypothesis:

**H2: After COVID-19, non-Chinese Asian restaurants will also experience a substantial drop in visits relative to non-Asian restaurants.**

Of note, these portrayals were not spread evenly across the population — Republican leaning news outlets, such as Fox News, were more likely to blame China for the pandemic than were other outlets (Krupenkin et al. 2021). Due to the prevalence of these narratives among Republican politicians and news sources, we argue that this shock was not homogeneously distributed throughout the population. Scholars have long noted that partisans in the U.S. tend to follow opinion cues provided by leaders of their political party (Druckman 2001; Lenz 2013; Bullock 2020). Partisans are much more likely to follow suggestions and absorb information when it is presented to them by leaders in their own party, as compared to non-partisan figures and especially figures from the opposing party. This partisan cuing effect extends to a variety of topics, including vaccination (Weisel 2021), mask-wearing (Milosh et al. 2021), and COVID-19 blame attribution (Graham and Singh 2021). This suggests that Republicans, whose party leaders and media outlets engaged in anti-Asian cuing behavior, should be more likely to
respond with consumer discrimination against Chinese and Asian restaurants, leading to Hypothesis 3:

**H3: Areas with more Trump support would see a larger relative drop in Chinese restaurant visits than areas with lower Trump support.**

Finally, we propose a potential mechanism for spillovers to non-Chinese Asian restaurants (such as “Tiger Belly Noodle Bar”) whose names are ethnically ambiguous but could potentially be classified as Chinese. During the pandemic, anti-China rhetoric often invoked prejudice and stereotypes such as the eating of bats or snakes and martial arts, which, while targeted at China, are common stereotypes of East Asians as a whole. Politicians’ usage of these stereotypes could both activate in-group favoritism and be indicative of the prevalence of these out-group stereotypes among their constituents. In either case, an activation of group membership has been shown to change how individuals process and perceive information about the out-group members, with the potential to increase perceived out-group homogeneity (Ostrom et al. 1993; Mullen and Hu 1989; Krosch et al. 2013). We thus predict that Trump voters who are more exposed to anti-China rhetoric from Republican leaning news outlets will exhibit decreased differentiation between Asian ethnic groups and thereby increased chances of misidentification. This yields our final hypothesis:

**H4: Trump voters would be more likely to misidentify non-Chinese Asian restaurants than non-Trump voters.**

We explore these hypotheses within consumer mobility data by comparing pre- and post-pandemic consumption behavior, supplementing our analysis with search data and a series of surveys to describe linkages between political rhetoric, sentiment, search, and subsequent behavior.

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8 John Cornyn criticized Chinese for eating snakes. He forgot about the rattlesnake roundups back in Texas. Last accessed May 15, 2022. https://www.washingtonpost.com/nation/2020/03/19/coronavirus-china-cornyn-blame/

9 President Trump calls coronavirus ‘kung flu’. Last accessed May 15, 2022. https://www.bbc.com/news/av/world-us-canada-53173436
Table 1: Data Summary

| Dataset                      | Measure                                      | Dates                                |
|------------------------------|----------------------------------------------|--------------------------------------|
| Google Trends                | Anti-Asian Attitudes and Food Safety         | 1/16 - 8/20                         |
| SafeGraph Mobility Data     | Visits to Restaurants                        | 11/4/2019 - 6/7/2020                 |
| Pollfish Survey Data        | Anti-Asian Attitudes                         | Fielded 5/5/20, 8/27/20, 10/29/20, 4/15/21 |
| Confusion Survey Data       | Restaurant Ethnic Misidentification          | Fielded 3/27/21-5/8/21               |

3 Data

We use four main data sources to determine the relationship between COVID-related anti-Asian sentiment and consumer discrimination against restaurants. Table 1 summarizes these data sources.

First, we used the web search data from Google to highlight the nature of the exogenous shock to American conceptions of China and Chinese food during COVID-19. Next, we use movement data provided by SafeGraph to directly measure consumer discrimination in the form of visits to Chinese restaurants, other Asian restaurants, and non-Asian restaurants. We then turn to understanding the underlying psychological mechanisms behind these forms of discrimination. We gathered survey data from 4,000 Americans (in four waves) to determine whether they blamed Chinese and other Asian American populations for COVID-19, and how this blame related to their beliefs about Chinese food safety. Finally, we use a survey with virtual workers to understand whether the spillover effects were driven by ethnic misidentification of other Asian restaurants as Chinese restaurants.
Search Data

We use weekly search data from US searchers via Google Trends in order to measure the effect of the COVID-19 pandemic as an exogenous shock to anti-Asian sentiment during 2020. Google Trends data has been shown to be a powerful measure of racial animus (Stephens-Davidowitz 2014) that correlates strongly with a wide variety of outcomes, including electoral (Stephens-Davidowitz 2014) and health outcomes (Chae et al. 2015).

Table 2 describes the search queries that we examine. We use search data for two main purposes. The first is to establish the temporal relationship between the COVID-19 pandemic and anti-Asian attitudes. Search data is uniquely well-suited for this purpose, as it measures organic increases in collective attention to a specific topic. To do this, we examine anti-Asian web searches that are not directly related to COVID-19. The reason is that while it is unsurprising that searches that tie COVID-19 to China would spike during the pandemic (so a search for “china virus” would not necessarily tell us anything about the searcher’s attitudes), searches for something like “china owns us” would be more indicative, since they have no clear relationship to the pandemic.

We look at several categories of anti-China searches: searches relating to China and Communism (political), searches relating to China and debt (economic), and searches relating to explicit anti-China phrases (social/stereotypical). All three of these categories are related to negative attitudes towards China and stereotypes about Chinese people and culture that predate the pandemic. We also compare searches for Chinese consumption of bats versus consumption of dogs. Search interest in Chinese bat consumption was likely driven by media coverage, but consumption of dogs has no relationship to the pandemic and did not receive media coverage. As such, if we see an increase in searches for Chinese consumption of dogs, this is another possible indicator of affirming the negative stereotypes about China.

The second purpose of using search data is to determine whether Americans treated food
| Category       | Type     | Google Trends Query                                                                 |
|----------------|----------|--------------------------------------------------------------------------------------|
| Anti-China     | Communism| china communist + chinese communist + china ccp + chinese ccp                         |
| Anti-China     | Debt     | china debt + chinese debt + china owns + owe china                                     |
| Anti-China     | Explicit | fuck china + boycott china + china sucks                                             |
| Anti-China     | Eating Bats| china eat bat + china eat bats + chinese eat bat + chinese eat bats                  |
| Anti-China     | Eating Dogs| china eat dog + china eat dogs + chinese eat dog + chinese eat dogs                  |
| Food Safety    | Restaurant Safety| food safe + takeout safe + restaurant safe + delivery safe            |
| Food Safety    | Chinese Restaurant Safety| chinese food safe + chinese takeout safe + chinese restaurant safe + chinese delivery safe |
isolate data from 381,692 unique restaurant locations. We focus on restaurants specifically due to their ubiquity, ease of affiliation between the business and a particular ethnic group, the rise of food safety-related searches following COVID-19, and historical and contemporary usage of food stereotypes as an outlet for xenophobic sentiment. We conduct our analysis within a difference-in-difference framework, measuring the differential impact of the onset of COVID-19 on visits to Asian versus non-Asian restaurants. This analysis controls for persistent differences in traffic for each restaurant and common temporal shocks.

**Attitudinal Survey Data**

We ran two samples of 1,000 respondents for the same set of questions conducted on May 5 and August 27, 2020 (noted as Survey 1), and slightly updated version on October 29, 2020 (Survey 2). We ran a final survey, also 1,000 respondents, on April 15, 2021 (Survey 3). The multi-waves were designed to capture the trends in our core questions of sentiment and behavior as the pandemic progressed. Surveys were conducted within native advertising units in a mix of mobile applications on both iOS and Android, and in both mobile and desktop web experiences, using the Pollfish interface and delivery mechanism. Participants were at least 18 years old, and randomly selected from the pool of eligible respondents. We asked a total of 25 questions in Survey 1 and 18 questions in Surveys 2 and 3 (see Appendix for a full list of questions that involved both exploratory questions as well as specific questions to be used in this paper). All survey results in the paper are raked to reflect national Census estimates of key demographics.

**Restaurant Confusion Survey Data**

We ran a separate study from March 26, 2021 through May 8, 2021 to examine in more depth the degree to which individuals distinguish between various Asian restaurants. 2,345 American
participants (51.3% female, age mean 39.4, age s.d. 12.6) who were eligible to vote in 2016 and 2020 were recruited through Amazon Mechanical Turk (MTurk). The 2,000 restaurants in the pool were chosen in a stratified random sampling from amongst mono-ethnic restaurants in the SafeGraph dataset. Each participant was asked to classify a randomly-selected set of 20 restaurants based on name alone from the pool of 2,000 real restaurants. Our stratification was chosen to balance between the most common ethnic restaurant cuisines while oversampling Chinese restaurants since they were the focus of messaging and food safety concerns. The sample included 400 Chinese restaurants and 200 each of French, Indian, Italian, Japanese, Korean, Mexican, Thai, and Vietnamese restaurants. Participants were also asked to self-report their gender, age, ethnicity, and voting preferences in both the 2016 and 2020 American presidential election.

4 Methodology and Results

Web Searches Confirm Exogenous Shock to Anti-Asian Sentiment

To examine the temporal shifts in the level of anti-Asian sentiment around the COVID-19 pandemic, we turn to the Google search data. Figure 1 presents the results of the first set of anti-China searches, which focus on non-COVID related anti-China sentiment. In all three cases, search patterns in 2020 looked very different from search patterns in the four years prior. For searches around debt and explicit anti-China searches, we see a sharp discontinuity at the week of March 11, 2020 that was not present in any of the prior years. In the case of Communism searchers, we see a slightly subtler effect, where the overall number of searches for China and Communism was substantially higher throughout the year, beginning in mid-March.

Searches related to eating animals, presented in Figure 2, further underscore these results. While there were clear, news-driven spikes for eating bats around the emergence of COVID-19

\[10\] This is despite the U.S. starting a well-covered trade war with China in 2019.
Notes: Google Trends shows sharp increases in negative searches relating to China during the week of March 11, 2020. There were no similar increases in these searches during the same time period in 2016-2019.

and again at the beginning of the pandemic in the U.S., the relationship between searches for eating bats and eating dogs is very clear and distinct. Dog-eating behavior had no relationship to the pandemic, but is closely related to anti-Chinese stereotypes.

These search patterns show a sharp discontinuity in anti-Chinese sentiment during the COVID-19 pandemic. Next, we will consider the impacts of this shift in sentiment on the livelihoods of Asian Americans in the U.S.

Asian Restaurants Experience Relative Drop in Visits During Pandemic

We analyze mobile device location data to document the direct impact of anti-China and anti-Asian sentiment on traffic to Asian American businesses in the post-pandemic period. This naturally lends itself to a difference-in-difference specification to capture the differential impact of the pandemic on traffic to Asian versus non-Asian restaurants. Specifically, we utilize the
Notes: Google Trends shows sharp increases in searches relating to Chinese people eating bats and dogs during the week of March 11, 2020. There were no similar increases in these searches during the same time period in 2016-2019. There is a substantial correlation between searches for eating bats and eating dogs, despite the fact that the dog-eating stereotype had no relevance to the COVID-19 pandemic.

following equation:

$$logVisits_{it} = \beta_1 \cdot I(Asian_i \times postCOVID_t) + \beta_2 \cdot caseRate_{jt} + \eta_i + \delta_t + \epsilon_{it}$$  \hspace{1cm} (1)$$

In this model, $i$ indexes restaurant locations and $t$ indexes the weeks from November 4, 2019 through June 7, 2020. $logVisits_{it}$ is the natural log of the visit count to restaurant $i$ in week $t$, and $caseRate_{jt}$ is the number of per capita reported cases of COVID-19 in week $t$ for the ZIP Code $j$ containing the restaurant $i$. Our difference-in-difference measure of interest is $\beta_1$, which captures the interaction between whether the restaurant $i$ is classified as an Asian restaurant and whether the time $t$ is after March 13, 2020, the major onset of COVID-19 in the United States and declaration of National Emergency. $\eta_i$ denote restaurant-level fixed effects capturing persistent differences in traffic for each restaurant location. These $\eta_i$ collectively capture differences between Asian and non-Asian restaurants, and thus replace the need
Table 3: Pandemic Impacts to Asian Restaurant Traffic

|                       | (1)                          | (2)                          | (3)                          | (4)                          |
|-----------------------|------------------------------|------------------------------|------------------------------|------------------------------|
| **Dependent variable:** | **log_visits**                |                              |                              |                              |
| isAsian x postCOVID   | $-0.203^{***}$               | $-0.115^{***}$               | $-0.115^{***}$               | $-0.288^{***}$               |
|                       | (0.015)                      | (0.014)                      | (0.014)                      | (0.0197)                     |
| isChinese x postCOVID |                              | $-0.288^{***}$               |                              |                              |
|                       |                              | (0.0197)                     |                              |                              |
| isNonChineseAsian x postCOVID |                      | $-0.288^{***}$               |                              |                              |
|                       |                              | (0.0197)                     |                              |                              |
| Case Rate             | $-7.224^{***}$               | $-7.500^{***}$               | $-7.703^{***}$               | $-7.296^{***}$               |
|                       | (1.383)                      | (1.395)                      | (1.455)                      | (1.397)                      |
| Week Fixed Effects    | Yes                          | Yes                          | Yes                          | Yes                          |
| Restaurant Fixed Effects | Yes                        | Yes                          | Yes                          | Yes                          |
| Observations          | 11,086,220                   | 10,256,567                   | 10,294,852                   | 11,086,220                   |
| R²                    | 0.837                        | 0.837                        | 0.837                        | 0.837                        |
| Adjusted R²           | 0.831                        | 0.832                        | 0.832                        | 0.831                        |
| Residual Std. Error   | 0.448 (df = 10728568)        | 0.448 (df = 9925678)         | 0.449 (df = 9962728)         | 0.448 (df = 10728567)        |

*p<.1; **p<.05; ***p<.01

Notes: Results of a difference-in-difference regression examining the relative drop in traffic for Asian versus non-Asian restaurants in the post-pandemic period. Week and restaurant fixed effects are utilized to account for parallel time trends and group membership, respectively. Robust standard errors are clustered on the week and restaurant level. Model (4) utilizes the same unrestricted dataset as Model (1), breaking the isAsian x postCOVID interaction into two separate interaction terms for Chinese and non-Chinese Asian restaurants in the post-pandemic period. Models (2) and (3) exclude non-Chinese Asian and Chinese restaurants, respectively, to ensure the comparison group is always non-Asian restaurants. Results show a significant drop in traffic for Asian, Chinese, or non-Chinese Asian restaurants in the post-pandemic period relative to non-Asian restaurants.

for an indicator on Asian restaurants in the typical difference-in-difference specification. δ capture common temporal shocks to traffic, such as if outdoor dining restrictions impacted all restaurants post-COVID or there was a reduction in traffic due to national holiday. As additional robustness, we also estimate models with state- and ZIP Code level fixed effects replacing restaurant fixed effects and state-week level fixed effects, which can capture regional variation in the onset and severity in COVID-19.

These results are summarized in Table 3. This analysis shows that in the period after the onset of COVID-19, Asian restaurant traffic dropped a substantial 18.4% (p<.001, 95% CI:
-15.9% to -20.8%) relative to non-Asian restaurant traffic in the same period and after controlling for COVID-19 case rates. Chinese restaurant traffic dropped 10.9% (p<.001, 95% CI: -8.4% to -13.3%), while non-Chinese Asian restaurant traffic dropped an even larger 25.0% (p<.001, 95% CI: -22.1% to -27.9%). These findings provide strong support for Hypotheses 1 and 2, which predicted both Chinese and non-Chinese Asian restaurants would experience substantial drops in visits relative to other types of restaurants.

We note that this difference-in-differences event study analysis measures the relative drop in traffic for Asian vs non-Asian restaurants during the pandemic, but on its own cannot distinguish between whether Asian restaurants are deciding to limit operations in anticipation of lowered demand or consumers are avoiding Asian businesses and thereby forcing a reduction in operations. That said, a mechanism whereby our observed drop in traffic is purely driven by a reduction in Asian restaurant operations is unlikely. Small business owners look to garner revenue to support themselves, their employees, and cover fixed expenses such as rent, suggesting that a greater reduction in Asian restaurant operations relative to comparable non-Asian restaurants absent changes in Asian restaurant demand would lead to unnecessary economic strain. Further, as we will demonstrate through our survey analysis, a sizable fraction of consumers expressed blame towards Asians for pandemic spread and fear that consuming even takeout Chinese food would present an increased risk of contracting COVID-19, suggesting the major role of shifting consumer demand for Asian food. And lastly, our study ultimately looks to quantify economic impacts of COVID-19 pandemic and associated stigmatization on Asian American restaurants and corresponding communities. These impacts would be felt similarly under both of these mechanisms.
Notes: ZIP Code level impact to Asian, Chinese, and non-Chinese Asian restaurants, respectively, by Trump voting share in 2016. Impact coefficients are calculated for each ZIP Code via difference-in-difference using the specification in Equation 1 relative to non-Asian restaurants within the ZIP Code. The grey shaded region displays the 95% confidence interval of the mean impact coefficient (blue line) of ZIP Codes with the given level of Trump support, visualized using LOESS smoothing with span = 0.5. These plots indicate that Trump voting share in 2016 correlates with greater avoidance of Asian, Chinese, and non-Chinese Asian restaurants in the post-pandemic period.

Greater Trump Support Associated with Larger Drop in Asian Restaurant Traffic

Given that former President Trump was a main vector of anti-China rhetoric assigning pandemic blame to China, we look to examine whether greater support for Trump was associated with a larger relative drop in Chinese and more broadly Asian restaurant traffic in the post-pandemic period. For this analysis, we utilize Trump’s 2016 voting share rather than his 2020 voting share, as his share in 2020 may have been impacted by both regional variation in pandemic severity and perceptions of his management of the pandemic.\(^{11}\) We explore heterogeneity in Asian restaurant avoidance by repeating our main analysis at the ZIP Code level, calculating the aforementioned difference-in-difference interaction effects for 1) all Asian restaurants, 2) Chinese restaurants, and 3) non-Chinese Asian restaurants, with results visualized in

\(^{11}\)As additional robustness, we replicate these results using the 2018 elections data in Appendix: 2018 Congressional Midterm Elections
Figure 3, and interaction regression results reported in Table 4. The ZIP Code level subsamples replicate our main effects, indicating persistent negative effects of the pandemic on Asian restaurants. Further, there is a visible negative correlation between traffic and Trump’s 2016 voting share at the ZIP Code level. This interaction is significant for each of Asian, Chinese, and non-Chinese Asian restaurants (p<.01) as shown in Table 4, which regresses the coefficient of avoidance at the ZIP Code level on each ZIP Code’s voting share in 2016. Lastly, while the average coefficients estimated in Table 3 showed greater avoidance of non-Chinese Asian restaurants than Chinese restaurants, we see in Figure 3 that this trend reverses for ZIP Codes with high Trump support. This provides support for Hypothesis 3, that greater Trump support would be associated with greater avoidance of Chinese restaurants.

The result that Chinese restaurant avoidance increased with Trump voting share supports that assignment of pandemic blame had downstream impacts on consumers’ food consumption choices. Given that a significant fraction of Asian restaurants in the U.S. serve Chinese food, the finding that Asian restaurants overall also saw a decline in traffic is unsurprising. What is less expected is that non-Chinese Asian restaurants saw a larger average decline in traffic than Chinese restaurants on a nationwide level.

In understanding this heterogeneity in effects between Chinese and non-Chinese Asian restaurants, we again look to Figure 3 and observe that ZIP Codes with very low levels of Trump support see lesser negative impact in traffic to Chinese restaurants relative to non-Asian restaurants in similar locations. Furthermore, the same figure shows that avoidance of Chinese restaurants is more sensitive to Trump support than avoidance of non-Chinese Asian restaurants\[12\]. We believe this is a reflection of how anti-Chinese sentiment became polarized during the pandemic. While former president Trump had utilized stigmatized language to refer to the pandemic, his political opponents such as then Speaker of the House Nancy Pelosi made efforts to explicitly support Chinatowns through the pandemic\[13\]. We also must recognize the

\[12\]This is explored further and reinforced by the results in Appendix: Three-Way Interaction Model

\[13\]Trump vs. Pelosi: What happened in Chinatown. Last accessed May 15, 2022. https://www.washingtonpost.com/politics/2020/04/23/trump-versus-pelosi-what-happened-chinatown/
Table 4: Heterogeneity in Effects

|  | Traffic Impact to Restaurant Type |  |
|---|---|---|
|  | Asian | Chinese | Non-Chinese Asian |
| Trump 2016 Vote Share | $-0.058^{***}$ | $-0.051^{***}$ | $-0.083^{***}$ |
|  | (0.011) | (0.012) | (0.014) |
| Constant | $-0.027^{***}$ | $-0.029^{***}$ | $-0.011$ |
|  | (0.005) | (0.006) | (0.007) |

Controls

|  |  |
|---|---|
| Median Household Income | Yes |
| College Education | Yes |
| Percent Population White | Yes |
| Percent Population Asian | Yes |

|  |  |
|---|---|
| Observations | 11,773 | 10,910 | 7,723 |
| R$^2$ | 0.002 | 0.002 | 0.004 |
| Adjusted R$^2$ | 0.002 | 0.002 | 0.004 |
| Residual Std. Error | 0.200 (df = 11771) | 0.208 (df = 10908) | 0.204 (df = 7721) |
| F Statistic | 28.766*** (df = 1; 11771) | 19.083*** (df = 1; 10908) | 34.523*** (df = 1; 7721) |

Notes: Heterogeneity in relative impacts to Asian, Chinese, and non-Chinese Asian restaurants. Impact coefficients are calculated for each ZIP Code via difference-in-difference via the specification in Table 3 relative to non-Asian restaurants within the ZIP Code and then regressed on the Trump 2016 vote share within the ZIP Code, controlling for median household income, college education, and racial demographics within each ZIP Code. The coefficient on Trump 2016 vote share is negative and significant in each case, reinforcing the correlation between Trump support and avoidance of Asian restaurants shown in Figure 3.

community efforts of Asian American organizations such as Welcome to Chinatown, Send Chinatown Love, and Think!Chinatown among many more through the pandemic to support Chinatowns, Chinese American restaurants, and Asian American businesses more broadly. Additionally, mainstream press coverage tended to focus on how Chinese American businesses and Chinatowns struggled through the pandemic, which, in combination with the community sup-

\[14\] Examples last accessed May 15, 2022:
https://www.fox5ny.com/news/chinatowns-pandemic-crisis-continues
https://www.wsj.com/articles/new-yorks-chinatown-businesses-struggle-to-survive-coronavirus-shutdown-11588856400
https://nypost.com/2020/07/16/nom-wah-used-frozen-dumplings-to-offset-coronavirus-struggles/
https://www.csmonitor.com/Business/2020/0310/When-fear-goes-viral-Battling-prejudice-in-Chinatown
port organizations described, likely led sympathetic consumers to specifically identify and support Chinese restaurants during this period. Asian American advocacy groups and efforts tend to be centered around urban areas and more liberal-leaning population centers, which shows up in our data as lower avoidance in low Trump support areas and affects our observed patterns on average avoidance of Chinese vs non-Chinese Asian restaurants nationwide.

On the other hand, we observe an inflection point when Trump support crosses around 65% and avoidance of Chinese restaurants is greater than avoidance of non-Chinese Asian restaurants. This is consistent with our survey results in the subsequent section which indicate correlations between political affiliation, pandemic blame towards Asian Americans, increased perceptions of risk from Chinese and Japanese food, and indications of outgroup homogeneity (such as overestimating the proportion of Asian Americans that are Chinese). In the locations where our surveys predict the greatest stigma against Asians, Asian Americans, and Chinese food, we do see greater avoidance of Chinese restaurants than non-Chinese Asian restaurants, and further these levels of avoidance are very high relative to comparable non-Asian restaurants in these areas. For these reasons, we believe that the patterns of observed heterogeneity reinforce rather than diminish the role of political racial stigma in driving consumption.

**Consumer Discrimination as an Underlying Mechanism**

**Surveys Show Blame on Asians and Fear of Chinese Food**

We start with a very straightforward question about blame: “Which racial or ethnic group do you believe is most responsible for bringing COVID-19 (aka coronavirus) into the U.S. (if any)?” For the first three waves we had one iteration of the question with six possible answers presented in a randomized order: "Asians", "Blacks", "Latinos", "Whites", "No racial

https://www.nbcnews.com/news/asian-america/year-loss-heartache-loomings-eviction-rescuing-itself-chinatown-n1249556
or ethnic group is responsible", and "Don’t know". We would expect, if anything, social desirability to motivate people to over-respond "No racial or ethnic group is responsible", but that answer hovers in the low 60’s, with a meaningful (but monotonically decreasing) 27% (Wave 1) to 17% (Wave 4) reporting "Asians" as the most popular ethnic group blamed. In Wave 4 on April 15, 2021 we randomly assigned half the sample to see a different set of answers to the same question, again in a randomized order: "Chinese", "Japanese", "Italian", "French", "Mexicans", "Greeks", "Indians", "Thai", "Other", "No racial or ethnic group is responsible", and "Don’t know". Even at this late date 38% answered "Chinese". While "No racial or ethnic group is responsible" is the most popular answer under all answer sets and days, a non-negligible amount of Americans blame Asians, and Chinese in-particular, for bringing COVID-19 into the U.S. The results are visualized in Figure 4.

Trump voters blamed Asians, in particular Chinese, at much higher rates than Biden voters. In the final survey on April 15, 2021 just 12% of Biden voters said they blame Asians, with 32% blaming Chinese. But, in the same survey 27% of Trump voters blamed Asians, and a majority, 56% blamed Chinese.

We also asked a direct question about fear and risk of contracting COVID-19: "When considering delivery food, ordering which (if any) of the following types of food do you believe presents an increased risk of contracting COVID-19 (aka coronavirus) into the U.S. (if any)?" This question allowed for multiple answers, and "Chinese food" was very consistently the leading answer with between 15% and 22%. This was between 5 to 9 percentage points higher than the next answer. Interestingly, Biden and Trump voters did not show much difference to this fear question, while our restaurant traffic analysis indicated that avoidance of Chinese and more broadly Asian restaurants is strongly correlated with support for Trump. This suggests that blame sentiment may have played a larger role for restaurant avoidance than fear of contracting COVID-19 from Chinese food.

Finally, we looked to examine the interaction between perceptions of Asian homogeneity
and directed blame or fear, as illustrated in Figure 5. We asked respondents, "What fraction of Asians in the U.S. do you think are ethnically Chinese?" We note two trends: first, that respondents consistently overestimated the fraction of Asians Americans that were Chinese (mean survey response 35% versus 24% actual\textsuperscript{15}). This is consistent with the predictions of out-group homogeneity bias, where respondents will tend to overestimate the homogeneity of the out-group. Second, examining respondents who believed that a majority (50% or more) of Asians in the U.S. were ethnically Chinese, we find this group is also significantly more likely to believe that Chinese food represented an increased risk of contracting COVID-19 (28% versus 17% among all respondents, two-sample Z-test: \( p < .01 \)) or that Asians were to blame for the

\textsuperscript{15}https://www.pewresearch.org/fact-tank/2021/04/29/key-facts-about-asian-americans/.
Figure 5: Fear, Blame, and Ethnicity Perceptions of Asian Americans

Notes: Responses to questions on fear and blame broken down by perceived percentage of Asian Americans that are ethnically Chinese from Waves 1 and 2 of the survey (May 5, 2020 and Aug 27, 2020, respectively). Respondents who perceive more Asian Americans are Chinese were also more likely to blame Asians for the spread of COVID-19 in the U.S. and believed Chinese food posed an increased risk of contracting COVID-19.

spread of COVID-19 (32% versus 22% among all respondents, two-sample Z-test: p < .01).

This second trend supports the relationship between perceived homogeneity of Asians as a group and fear and blame directed towards Asian Americans.

Increased Searches Relating to Chinese Food Safety

To delve further on the question of fear versus discrimination, we examine the search patterns around Chinese restaurants and other restaurants from 2016-2020. Figure 6 shows that searchers exhibited very different search patterns around Chinese restaurants. For regular searches around restaurant safety, we see a large spike in the week immediately following Trump’s March 11 address. For Chinese food safety searches, however, the largest spikes came in Febru-
Figure 6: Food Safety Searches

**Restaurants Safe**

**Chinese Restaurants Safe**

|  |  |  |  |
|---|---|---|---|
| 2016 | 2017 | 2018 | 2019 | 2020 |
| Jan | Feb | Mar | Apr | May | Jun |

Notes: Google Trends data shows that searches relating to concern about the safety of Chinese food peaked earlier than searches relating to concerns about the safety of restaurants in general.

ary and early March 2020, with the largest spike during the week of March 11. These results show that people were thinking about the safety of Chinese restaurants and other restaurants differently during the pandemic.

**Confusion Among Asian Restaurants a Potential Mechanism for Spillover Effects**

Following on our mobility analysis, we investigate the potential for anti-China messaging spillovers to create a decline in business for non-Chinese Asian restaurants. We specifically examine confusion among Asian restaurants as a potential mechanism, motivated by extant literature on out-group homogeneity bias and a history of attacks against Asian Americans where attackers were motivated by hatred against one Asian group yet attacked victims of another Asian group.$^{16}$

16Vincent Chin murder 35 years later: History repeating itself?. https://www.freep.com/story/news/2017/06/24/murder-vincent-chin-35-years-ago-remembered-asian-americans/420354001/. Retrieved May 6, 2021;

Report: Sam’s Club stabbing suspect thought family was ‘Chinese infecting people with coronavirus’. https://www.kxan.com/news/crime/report-sams-club-stabbing-suspect-thought-family-was-chinese-infecting-people-with-coronavirus/. Retrieved May 6, 2021;

Texas ramen shop still requiring masks hit with anti-Asian graffiti. https://www.nbcnews.com/news/asian-america/texas-ramen-shop-still-requiring-masks-hit-anti-asian-graffiti-n1261129. Retrieved May 6, 2021.
Notes: Misidentification rate of Asian restaurants split based on voting preferences. MTurk survey conducted on 2,345 participants (51.3% female, age mean 39.4, age s.d. 12.6) asked to label a randomly selected sample of real, mono-ethnic restaurants from the SafeGraph dataset by ethnic group. p-values shown are results of a two-sided z-score test for difference of proportions.

We conducted a survey to test this mechanism with 2,345 participants. Each participant was shown a list of 20 ethnic restaurant names (selected from Chinese, Japanese, Korean, Thai, Vietnamese, Indian, Mexican, Italian, and French), randomly selected from real, mono-ethnic restaurants within the SafeGraph location dataset, and asked to categorize each one by its ethnicity. We find that misidentification is high: 33% of Asian restaurants are incorrectly labeled by respondents. The most commonly mislabeled Asian restaurants were Korean and Vietnamese, which were incorrectly labeled 52% and 47% of the time, respectively. Both of these were most commonly mislabeled as being Chinese, consistent with a misclassification mechanism behind Asian restaurant avoidance.

We also break down correct response rate by participants’ 2016 and 2020 voting prefer-
ences, as shown in Figure 7. These results indicate that Trump 2016 and 2020 voters were more likely to misidentify Asian restaurants (36.4% and 35.2% misidentification, respectively) compared to Clinton 2016 and Biden 2020 voters (30.9% and 31.6% misidentification, respectively). Two-sample Z-tests for a difference of proportions comparing Trump vs. Clinton voters in 2016 and Trump vs. Biden voters in 2020 indicate these differences in misidentification rate are highly significant (p<.001 for each).

Recalling our other survey results, we noted a significant interaction in that respondents who (falsely) believed that a majority of Asians in the U.S. were Chinese were also more likely to report a fear that Chinese food represented an increased risk of contracting COVID-19. We also find that greater Trump support is associated with both increased misclassification of Asian restaurants and greater avoidance of non-Chinese Asian restaurants (Figure 3). Taken together, these results illustrate the linkage between expressions of out-group homogeneity bias (perceived increased homophily or inability to distinguish between out-groups) and Asian restaurant avoidance attitudes and behaviors. These patterns are all highly consistent with a mechanism whereby non-Chinese Asian restaurant avoidance is influenced by ethnic misidentification as stated in Hypothesis 4. We caution, however, that this does not constitute a causal test of the role of ethnic misidentification or allow us to rule out alternative mechanisms arising from out-group homogeneity such as collective blame, whereby a consumer harboring blame towards Asians for causing the pandemic may avoid Korean restaurants due to the broad association of Korean cuisine with Asia without directly needing to confuse Korean and Chinese cuisines.

5 Conclusion

In this paper, we examined the effects of anti-Asian sentiment on consumer discrimination against restaurants associated with Asian Americans. Leveraging and documenting the COVID-
19 pandemic as an exogenous shock to anti-Asian sentiment, we find a marked and substantial decline in visits to both Chinese and non-Chinese Asian restaurants. In line with partisan cu-
ing theories, we find this effect to be most prominent in areas with more Trump voters. While Trump voters were not more likely to express safety concerns with consumption of Chinese food, they were significantly more likely to attribute blame for pandemic spread to Asians or Chinese. This suggests that anti-Asian sentiment drove restaurant avoidance beyond perceived risk of contracting COVID-19. Finally, we examined a potential driver of spillovers to non-Chinese Asian restaurants in ethnic misidentification. Consistent with our hypothesis, survey responses showed that individuals who incorrectly believed the majority of Asian Americans were Chinese also expressed greater fear of Chinese food, and Trump support correlated with both increased avoidance of non-Chinese Asian restaurants and ethnic misidentification of Asian restaurants.

Given our findings, it is unlikely that legitimate concerns over health from consuming restaurant food explain the disproportionate drop in visits for Chinese restaurants as well as non-Chinese Asian restaurants. Rather, it is likely that anti-China sentiment – regardless of whether it was driven by fear or anger – triggered by COVID-19 and reinforced by U.S. polit-
cical rhetoric played a significant role in consumer discrimination against Asian-owned restau-

It should be noted that concerns about rising anti-Asian discrimination extend far beyond the consequences of the COVID-19 pandemic. Asian Americans have made up the plurality of immigrants to the U.S. since 2008 (Budiman2020). Historically, the large increases in California’s Chinese immigrant population during the 19th century were swiftly followed by a sharp increase in anti-Asian sentiment and discrimination, which persisted even after legisla-
tion stopped the flow of new immigrants from Asia (Liu2015, Chiu and Kirk2014). In addi-
tion to this historical example, recent research (Cikara, Fouka and Tabellini2021) has found that as minority groups increase in rank in terms of their relative size (e.g., to largest minority
within a community), negative attitudes and violence toward these minority groups increase as well. Together with our own findings, these examples suggest that the U.S. may be entering a longer period of increased anti-Asian sentiment, which makes understanding the effects of discrimination against this group even more important.

A number of limitations of this research derive from the fact that the onset of COVID-19 was a single, unprecedented shock to public health and the global economy. For example, we cannot entirely disentangle the relative role of political leaders’ anti-Asian rhetoric from endogenous anti-Asian sentiment arising from the public health threat itself. History has repeatedly shown that blaming marginalized groups for public health crises is sadly quite common (Perry and Donini-Lenhoff 2010; Keusch, Wilentz and Kleinman 2006). For example, Irish immigrants were blamed as the bearers of cholera, and Jewish immigrants were blamed for tuberculosis (Kraut 2010). More recently, African and Chinese cultural habits were also blamed for the outbreaks of Ebola and SARS (Kapiriri and Ross 2020), and Latinos were blamed for the spread of H1N1 (McCauley, Minsky and Viswanath 2013). It is possible that anti-Asian sentiment would have arisen even absent public officials’ statements targeting Asians, or that politicians’ statements are reflective of constituent beliefs rather than formative. However, the long history of stigmatization, discrimination, and violence toward minority groups following disease outbreaks should put greater onus on public officials to avoid such statements due to exactly these risks to minority communities. While we cannot reliably create a counterfactual world where COVID-19 was not followed by anti-China rhetoric by the U.S. political leaders, future research might further examine the relative importance of different triggers for anti-Asian sentiment, discrimination, and behavior.

Ethics Oversight

The Institutional Review Board for Health Sciences and Behavioral Sciences at the University of Michigan has re-
viewed the two studies and determined that they were exempt from IRB review (Attitudinal survey data: HUM00180988; Restaurant confusion survey data: HUM00195836).

**Data and Code Availability**

Consumer mobility analysis code and survey data and analysis code which support the findings of this study have been deposited in an Open Science Framework repository accessible through this link: [https://osf.io/gazwb/?view_only=cafc11d21236433caa3781748416fa9d](https://osf.io/gazwb/?view_only=cafc11d21236433caa3781748416fa9d). Consumer mobility data utilized in this study are derived from the Weekly Patterns and Core Places datasets provided under a free academic license from SafeGraph, however license restrictions do not allow the datasets to be made publicly available. Lastly, this study utilizes publicly available datasets on Google search ([https://trends.google.com/trends/](https://trends.google.com/trends/)) COVID-19 cases ([https://www.nytimes.com/article/coronavirus-county-data-us.html](https://www.nytimes.com/article/coronavirus-county-data-us.html)), US Census information ([https://www.census.gov/data.html](https://www.census.gov/data.html)), and 2016 and 2020 US presidential election results ([https://electionlab.mit.edu/data](https://electionlab.mit.edu/data)).

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