

# Sexual Identity, Poverty, and Utilization of Government Services\*

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## Abstract

Previous literature has established that lesbian, gay, and bisexual (LGB) people are at least as likely to be poor as heterosexual people, standing in contrast to myths of “gay affluence.” These findings have used datasets limited by either sample size or using partnership status to infer sexual orientation. Using U.S. data from the Household Pulse Survey, which allows us to identify large samples of individuals who self-identify as lesbian, gay, or bisexual, we find that bisexuals have lower incomes, are more likely to experience poverty, and bisexual individuals, gay men, and lesbian women are more likely to report financial hardship. Additionally, we find that LGB people utilize government assistance at higher rates than heterosexual people, even when allowing for selection into poverty status. We propose several explanations for these differentials, drawing on the program non-participation literature, and suggest that social network effects, lessened stigma, and increased reliance on public programs may explain these differences. Finally, we examine receipt of the enhanced child tax credit and find evidence that gay men and lesbian women with children were less likely to receive it than heterosexual men and women with children.

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## 1. Introduction

A large body of research has documented that sexual minorities (e.g. lesbian women, gay men, and bisexual populations—henceforth referred to as LGB individuals) have worse economic outcomes than heterosexuals, with lower incomes, higher rates of poverty, lower rates of health insurance coverage, and lower rates of homeownership (Albelda et al., 2009; Badgett et al., 2021; Elton & Gonzales, 2022; Gonzales & Blewett, 2014; Leppel, 2007; Uhrig, 2015). Additionally, some literature has examined utilization of government assistance programs by sexual orientation, typically used as a proxy for economic distress (Brown et al., 2016; Everett & Mollborn, 2014; Schneebaum & Badgett, 2019).

Tangentially, there is also large literature in public economics examining the delivery, take-up, receipt, and utilization of public assistance programs (e.g., SNAP, Unemployment Insurance, etc.). Nearly all public assistance programs do not have perfect take-up among eligible recipients, and these rates can be crucial for projecting the aggregate costs and impacts of government assistance programs, as well as evaluating the welfare implications of policies that have already been implemented (Currie, 2004; Finkelstein & Hendren, 2020; van Oorschot, 1991). Several possible explanations have been proposed for the drivers of imperfect take-up, including low monetary gains, stigma of receipt, time costs associated with program participation, imperfect information, administrative barriers, and mismeasurement (Ko & Moffitt, 2022). Alternatively, it is possible that differences in access and take-up to government assistance programs can drive or amplify economic disparities experienced by marginalized populations. We open a dialogue between these subject areas and add novel evidence to the large body of public economic literature by identifying differential take-up of public assistance programs between LGB and heterosexual adults and proposing several explanations for future research to examine.

Understanding the economic position and receipt of government assistance among sexual minorities is also important for several reasons. First, sexual minorities comprise a large and increasing share of the American population, and sexual minorities have gained much-needed recognition in public policy. Yet, there are substantial and ongoing policy implications that can be informed by LGB differentials in government assistance take-up, given the magnitude of economic costs that could occur. Badgett et al., 2021 report several estimates in the United States, finding that 2.7-4.6 percent of the American adult population identifies as non-heterosexual. Recent analyses have found that 15-20 percent of high school students identify as non-heterosexual (Gonzales & Deal, 2022; B. Wilson & Meyer, 2021). These estimates imply that there may be large policy implications that proceed from LGB differentials in government assistance take-up, given the magnitude of costs that could occur. Second, some public assistance programs are designed to benefit certain demographic characteristics and family structures (Hoffman, 2008; Hussey, 2011). LGB people may differentially receive or utilize public assistance programs, given that their demographic characteristics differ dramatically from heterosexuals (Brewster et al., 2014). In this respect, our analysis points towards a growing literature on population-targeted

policies—such as the Affordable Care Act—and their differential effects on LGB individuals (C. S. Carpenter et al., 2021; C. S. Carpenter & Sansone, 2021; Marcén & Morales, 2022).<sup>4</sup>

Prior research on LGB poverty and government assistance receipt has been limited by small sample sizes and the necessity of using same-sex couple status to infer sexual orientation (which excludes single sexual minorities and bisexual individuals and may lead to unrepresentative estimates) (Albelda et al., 2009; Badgett et al., 2013; Badgett, 2018; DiBennardo & Gates, 2014; Schneebaum & Badgett, 2019; Uhrig, 2015). Related work examines LGB disparities in several key economic vulnerability measures during the pandemic, including pandemic-related job loss, food insufficiency, and housing insecurity, finding broad negative differentials in economic security for LGB adults (M. Martell & Roncolato, 2022). We leverage the large sample sizes of self-identified sexual minorities in the U.S. Census Bureau’s Household Pulse Survey (HPS) to evaluate the economic outcomes and government assistance receipt of sexual minorities relative to their heterosexual counterparts. We are among the first, to the best of our knowledge, to examine differences by sexual orientation across subjective measures of economic status. Additionally, we evaluate several government assistance programs that have not previously been examined in the context of differences by sexual orientation: the child tax credit, unemployment insurance, rental assistance, and stimulus payments. We also examine whether government assistance differentials for sexual minorities are homogeneous for bisexual and lesbian/gay individuals, identifying disparities that may face a subset of sexual minorities but be obscured in overall figures. Finally, we use data collected during the COVID-19 pandemic, a period of heightened economic distress when government assistance programs were increasingly utilized across the population. Evaluating the receipt and utilization of public assistance programs by sexual orientation is crucial for policymakers that must consider the potential for heterogeneous effects of policy on different communities, especially those with a history of worse economic outcomes.

We report several key results. We find that bisexuals have lower incomes and are more likely to report household incomes below the federal poverty level (FPL) than heterosexuals. Similarly, gay men, lesbian women, and bisexual individuals self-report higher economic insecurity than heterosexuals. Additionally, we find that sexual minorities are more likely to receive and utilize government services than heterosexuals. Differences in receiving government assistance persist when allowing for selection into FPL status, and we find evidence of higher utilization among gay and bisexual men compared to heterosexual men when restricting to a subsample of individuals who report household incomes below the FPL. We explore several potential reasons for this differential in our discussion, including social network effects, stigma differences surrounding the use of public assistance, and education differences. Finally, we evaluate receipt of the child tax credit by sexual orientation and find that gay men and lesbian women are less likely to receive the child tax credit than their heterosexual peers, even after adjusting for the presence of a child in the

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<sup>4</sup> Irrespective of these reasons, it is important to study the outcomes and economic wellbeing of marginalized populations regardless of group size or demographic characteristics, as LGB populations have historically been neglected in the economics literature (though the quantity of research has grown significantly) (Badgett, 1995; Badgett et al., 2021).

household and the total number of children in the household. This difference holds even for the subsample of individuals who report household incomes below the FPL.

The remainder of the paper takes the following form. Section 2 reviews previous literature on sexual minority economic status, as well as access to and utilization of government assistance programs. Section 3 develops a conceptual framework for LGB differentials in public assistance take-up and access. Section 4 describes our data and empirical approach. Section 5 presents the results, and Section 6 concludes.

## **2. Literature Review**

### *Poverty*

The U.S. Census Bureau measures poverty by comparing a person or family's household income to a threshold dependent on family size (Brady, 2005). One avenue through which the U.S. government addresses poverty is public assistance programs, including cash support, provision of necessities (e.g., health insurance and funding to subsidize food or housing), and employment programs.

Due to discrimination, structural racism, and historical inequities, minority groups are often disproportionately affected by poverty and have reduced access to public resources (Horsfall, 2012). For example, people with disabilities and Hispanic people are less likely to participate in programs like SNAP (Supplemental Nutrition Assistance Program), often because they do not know about benefits or are unable to navigate the enrollment process (Alvira-Hammond & Gennetian, 2015; Keith-Jennings et al., 2019).

The theoretical impact of the COVID-19 pandemic on poverty is unclear. While the pandemic certainly interrupted the economy, causing widespread unemployment and economic distress for families, the federal government responded with historic public assistance efforts like the advance CTC, Paycheck Protection Program, and pausing Medicaid disenrollment. This is observed in previous empirical analyses of economic wellbeing, which have broadly found that pandemic-related government policy effectively countered its effects on incomes, leading poverty to fall and the household income of poorer households to rise across a range of demographic groups and geographies (Han et al., 2020; M. Martell & Roncolato, 2022).

### *LGB Economic & Poverty Status*

Several economic studies have documented income differentials and penalties for sexual minorities. Gay and bisexual men have lower average incomes than do heterosexuals, whereas differentials for lesbian and bisexual women vary by study (Badgett et al., 2021; Drydakis, 2022; Plug & Berkhout, 2004). Earnings may be as much as 16% lower for gay men and 15% higher for lesbian women compared to their heterosexual counterparts (Ahmed & Hammarstedt, 2010; Klawitter, 2015); it is likely that the disparities observed for gay and bisexual men are due to taste-based discrimination, a finding that has been reinforced by audit studies (Badgett, 1995; Patacchini

et al., 2015; Tilcsik, 2011). Additionally, prior literature has indicated that certain subpopulations of the LGB community, especially bisexuals, are more likely to experience poverty than heterosexuals (Badgett, 2018). The same pattern does not hold, however, for lesbian women and gay men, who are less likely to experience poverty than heterosexuals (Badgett, 2018; Uhrig, 2015). This bisexual poverty differential may proceed from compositional differences, as bisexual men and women tend to be younger, have lower educational attainment, and are more likely to be never married and unpartnered compared to both their heterosexual and gay/lesbian counterparts, all characteristics associated with higher relative risk of poverty (Badgett, 2018). Additionally, we find that bisexual men and women are more likely to have children, increasing their relative risk of poverty. However, there is also a growing literature examining bisexual health that proposes a role for “double discrimination,” whereby bisexual individuals experience stigma and minority stress from both within and beyond the LGBTQ community (Colledge et al., 2015; Feinstein & Dyar, 2017). It is possible that this stressor may contribute to the unique disparities in economic outcomes that bisexual individuals experience.

Several factors may exacerbate these disparities. Gay and bisexual men without children experience higher poverty rates than heterosexuals without children and the children of same-sex couples are twice as likely to be poor than children of different-sex married couples (Albelda et al., 2009). Comparing across race and location yields discrepancies in poverty rates as well. African American same-sex couples are much more likely to experience poverty than white same-sex couples. LGB African Americans experience poverty at least twice as much as their heterosexual counterparts (Badgett et al., 2013). Finally, same-sex couples in rural areas have a poverty rate that is twice that of those in metropolitan areas (Albelda et al., 2009).

Beyond labor market outcomes, other factors can influence an individual’s wellbeing. Perceived financial insecurity and risk are alternative measures of economic health that may offer a more comprehensive account (Western et al., 2012). Negative labor market outcomes tend to spillover to overall wellbeing, suggesting that we may expect negative financial wellbeing differentials for LGB populations compared to heterosexuals. For example, American LGB populations are also disproportionally food insecure (Brown et al., 2016). Previous research has also found elevated levels of economic insecurity among sexual minority men (Chai & Maroto, 2020; Mann et al., 2019).

#### *LGB Government Assistance*

Less data is available on LGB access to government assistance due to the lack of questions regarding sexual orientation and gender identity in past surveys. One analysis suggests that same-sex couples—especially men in same-sex couples—access public health insurance at a higher rate (Badgett et al., 2006). Additionally, other analyses have found higher rates of cash assistance receipt among sexual minorities (Badgett et al., 2013; Uhrig, 2015). Prior studies typically use receipt of public assistance as a proxy for poverty status and economic well-being.

### **3. Conceptual Framework**

Economic modelling of non-participation in public assistance programs generally ascribes the behavior to some combination of stigma, imperfect information, access costs outweighing participation benefits, and social networks (Currie, 2004). For example, stigma was introduced first as a general disutility associated with program participation, and then this model was expanded to include a broader cost/benefit framework (Moffitt, 1983). Additionally, a large empirical and theoretical literature has documented and modelled social network effects in program participation (Aizer & Currie, 2004; Bertrand et al., 2000; Borjas & Hilton, 1996; Manski, 1993). Geographic and social networks can facilitate increased take up of public assistance programs. Information barriers can also keep eligible potential recipients from realizing their eligibility (Ko & Moffitt, 2022). We use these models of non-participation and prior work on the demographic characteristics of LGB populations to generate predictions for the public assistance takeup of this group.

Several demand-side mechanisms may lead to increased public assistance participation for LGB adults. LGB individuals are more likely to be in poverty than heterosexual individuals and are resultantly more likely to qualify, receive, and utilize government assistance to support their economic conditions. In fact, some previous studies have used receipt of government assistance as a proxy for poverty and economic distress (Badgett, 2018; Badgett et al., 2006; Uhrig, 2015).<sup>5</sup> In addition to pre-existing vulnerabilities, another possibility is that LGB individuals may have been differentially affected by the COVID-19 pandemic and so may be more likely to be unemployed and require government assistance as a result. Gonzales & Loret de Mola (2021) found that sexual minorities were more likely to be working in COVID-19 sensitive industries, which would support the hypothesis that they are more vulnerable to employment disruptions (and subsequently need more assistance) caused by the COVID-19 pandemic than otherwise similar heterosexuals. Additionally, Martell & Roncolato, 2022 found that lesbian women, bisexual women, and bisexual men were more likely than their heterosexual counterparts to be in a household that experienced pandemic related job loss. These demand-side explanations suggest that LGB people may have higher public assistance take-up due to a higher need for support.

However, we might also find differentially larger take-up of government assistance programs by LGB individuals compared to similarly situated heterosexual individuals. Why might this be the case?

Information barriers are a common explanation for program non-participation and can be mitigated by education (Currie, 2004). Sexual minority adults are more likely to have college and advanced degrees than heterosexuals (Black et al., 2007; C. Carpenter & Gates, 2008; Gonzales & Blewett, 2014). More educated individuals may be more likely to be aware of and navigate bureaucratic hurdles to government assistance programs, thus enabling them to take up the programs at higher rates by reducing information barriers and access costs. Now, we turn to two other possibilities: stigma and a lack of other support networks.

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<sup>5</sup> However, this could be misleading if there is differential take-up even conditioning on economic status.

Another factor that influences participation in government assistance programs is stigma surrounding the receipt and utilization of public assistance programs (Allen et al., 2014; Fothergill, 2003; Stuber & Schlesinger, 2006). Some individuals who qualify for benefits may not choose to use them due to political views or principles surrounding whether the government should provide such benefits (Morin et al., 2012). Ample social science research has found that sexual minorities tend to align with left-leaning parties and support liberal policy positions, suggesting that they may not share this stigma to the same degree as heterosexual populations (Edelman, 1992; Jones, 2021; Lewis et al., 2011; Smith & Haider-Markel, 2002; Turnbull-Dugarte & Townsley, 2020). The stigma hypothesis of higher support for and willingness to use government assistance programs suggests a positive public assistance take-up differential for sexual minorities.

A lack of alternative community-based and familial networks for low-income LGB people may indicate they have to rely on government assistance more than similarly situated heterosexuals. Previous research has established that LGB people are less likely to be affiliated with organized religion, which is one pathway through which individuals can acquire social and economic supports (Herek et al., 2010; Sherkat et al., 2010). Additionally, the family strains involved in the coming out process or a lack of family acceptance may lead to decreased social or economic support from immediate and extended family members, though some work has found higher financial transfers from parents to LGB children than heterosexual children (Dempsey et al., 2020; Perales & Huang, 2020). Government assistance may be filling in gaps for low-income LGB people where they lack support structures that low-income heterosexuals have, again predicting higher take-up among sexual minorities (Wilson et al., 2020).

Finally, it is possible that differentials are due to social network effects. Sociologists have established that individuals' choices can be influenced by their peers, and that these "social networks" can shape behavior both consciously and unconsciously (DiMaggio & Garip, 2012). Additionally, some literature has found that subgroups of the LGB community have denser (non-familial) social networks than heterosexuals (Breder & Bockting, 2022). It is possible that LGB individuals in poverty may be referring each other to social service access differentially, or have access to social service providers and assistance through networks like sexual health clinics for HIV and STI care, which frequently offer comprehensive services alongside sexual health screenings and treatments (AIDS Project LA, 2022). Additionally, individuals with HIV may automatically qualify for SSI and Medicaid if their health deteriorates to disabling levels.

Using the frameworks outlined above and the contextual details of the LGB community, we expect that LGB adults will take-up public assistance programs at higher rates than similarly eligible heterosexual adults.

## 4. Data/Methods

### 4.1 The Household Pulse Survey

Data for our study are drawn from waves 3.2-3.4 (weeks 34-43) of the Household Pulse Survey (HPS), a nationally representative and repeated cross-section of approximately 70,000 households in the 50 U.S. states and the District of Columbia. Respondents are surveyed in short waves that the US Census Bureau calls “weeks,” usually spanning 12-15 days and surveying approximately 70,000 households. We combine these weeks to improve statistical power and, after applying sample restrictions (outlined further below), we have a primary analytic sample of 523,796 respondents, including 19,007 gay or lesbian individuals, as well as 19,290 bisexual individuals. The HPS was designed to measure rapid responses to the COVID-19 pandemic; thus, the HPS is a rich source of information on how the pandemic affected health, finances, income security, and utilization of public assistance programs. These data were collected and made publicly available through the U.S. Census Bureau and contain demographic, economic, social, and housing information. As a large household survey administered by the U.S. Census Bureau in conjunction with other agencies, the large sample sizes of the HPS make it particularly germane to studying subpopulations like low-income individuals and/or the LGB population. Other researchers have used these data to examine the social safety net (Bitler et al., 2020), education supply (Bansak & Starr, 2021), and consumer behavior (Garner et al., 2020a).

The HPS is one of the only large national surveys that directly asks respondents about their sexual orientation. Respondents are given the prompt: “Which of the following best represents how you think of yourself?” and they can answer (1) Gay or lesbian; (2) Straight, that is not gay or lesbian; (3) Bisexual; (4) Something else; and (5) I don’t know. This question was added to the questionnaire in week 34, so we used data from weeks 34-43, which were fielded in August 2021 through March 2022. We focus on those respondents who indicated that they were (1) Gay or lesbian or (2) Bisexual, which we designate as sexual minorities for our analysis. This direct question ascertaining sexual orientation identity is preferred to inferring sexual orientation through same-sex couple status due to issues of representativeness and capture (C. S. Carpenter et al., 2021; M. E. Martell, 2021). The HPS also asks individuals about their gender identity using a two-step process. We use this information to control for gender identity but focus our analysis on differences by sexual orientation.<sup>6</sup> For more information on gender minority economic outcomes in the Household Pulse Survey, see (C. S. Carpenter et al., 2022).

Regarding economic outcomes, participants self-report their employment status and household income in ranges. Specifically, all individuals report whether they worked for pay or profit in the last seven days and their 2020 household income in ranges (Less than \$25,000; \$25,000 - \$34,999; \$35,000 - \$49,999; \$50,000 - \$74,999; \$75,000 - \$99,999; \$100,000 - \$149,999; \$150,000 - \$199,999; \$200,000 and above). We examine household income directly, as well as poverty status, which is based on household income and household size (individuals are asked to state the number

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<sup>6</sup> We exclude respondents whose sex assigned at birth was allocated by the HPS hot-decking procedure.



of adults and the number of children in the household; further information about the construction of this variable can be found in Appendix Table 2). The income-based thresholds used to determine the federal poverty level status of an individual come directly from the U.S. Census Bureau, where these thresholds were used to estimate the official poverty rate in the US (US Census Bureau, 2021). However, eligibility for programs administered by the Department of Health and Human Services, including Medicare, Medicaid, CHIP, SNAP, and TANF, is determined using the federal poverty guidelines (FPG), a slightly different definition of poverty (Office of the Assistant Secretary for Planning and Evaluation, 2022). Both calculations consider the household composition and income, but the FPL also considers elderly status in determining poverty. As a result, our estimates of FPL status do not perfectly capture the factors that determine eligibility for some government assistance programs. Additionally, because we are using the midpoint of the income category that respondents report to determine poverty status, there is potential for mismeasurement as respondents whose exact income (which is unobserved) falls above the midpoint of an income category might be miscoded as below the FPL if their exact income falls above the FPL threshold (and vice versa for respondents with incomes below the midpoint of their income category). Because income is generally monotonically decreasing over the income ranges we observe, it is likely that there are more individuals whose income lies below the midpoint of the income category we use (and may thus be miscoded as above the FPL) (Congressional Research Service, 2021). Therefore, it is likely that this measurement bias means we are underestimating the prevalence of poverty in our sample, though the magnitude of this bias is not observable. We also use an indicator variable for financial hardship, where respondents are asked whether it has been difficult for their household to pay for usual expenses in the past 7 days. Regarding public assistance receipt, participants in the HPS are also asked whether they or anyone in the household receives or has received food assistance benefits through the Supplemental Nutrition Assistance Program (SNAP) or a child tax credit in the past four weeks. Respondents are also asked about the source of their health insurance coverage, including Medicaid coverage for low-income families and individuals. Additionally, respondents were asked which of the following sources they and their household members used to meet their spending needs. They could choose multiple options including: unemployment insurance (UI), stimulus payments, SNAP, school meal cards, and governmental rental assistance. This distinction between receipt and utilization of government services, both self-reported, is necessary to understand why discrepancies between the two can emerge.

## **4.2 Data Quality and Limitations**

There are several limitations to the data from the HPS. First, about 18.8 percent of respondents did not provide a response to the household income question, which we use to determine poverty and FPL status.<sup>7</sup> This is common in surveys where participants are asked to provide information on

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<sup>7</sup> This rate is lower among gay men and lesbian women (14.2 percent) and bisexuals (17.4 percent) than heterosexuals (19.1 percent). This may suggest that our estimates are an upper bound of economic disparities, as respondents missing income data likely had worse economic outcomes.

their income (Bhat, 1994). This study only uses complete cases and information for all analyses to avoid imputations for missing data. We report the distribution of educational attainment for respondents who reported their income and for those who did not in Appendix Figure 1, which shows that our analytic sample (those who reported income) is positively selected on educational attainment relative to the non-respondents. This limits our ability to study these non-respondents, who may be accessing government services at higher rates than the rest of our sample, given their educational attainment. Several analyses have examined potential determinants of item non-response on population-based surveys and found that demographic and geographic characteristics like age, race, and region of residence are significantly associated with nonresponse to income and wealth questions while survey-specific characteristics like interviewer-interviewee concordance on socioeconomic status play a smaller role (Riphahn & Serfling, 2005; Schenker et al., 2006). For employment and other socioeconomic outcomes, missingness rates were very low. Additionally, respondents to the income question report answers in ranges rather than exact amounts. We used the midpoint of each increment to measure individual-level household incomes. Another limitation of the data is that sexual identity is self-reported, raising the potential for response bias. Approximately 2% of adults in the Household Pulse Survey said that they did not know how to respond to the sexual orientation question. Nonresponse to self-identified sexual orientation questions is associated with race and ethnicity, so this feature of the data may indicate that we do not fully capture sexual minorities, especially people of color (H.-J. Kim & Fredriksen-Goldsen, 2013). Moreover, very little research has been able to measure misreported data. Our findings are likely biased towards the null since economically disadvantaged populations are most likely to misreport or skip the sexual orientation question.

In terms of limitations, other researchers have raised concerns about the representativeness of the HPS data, especially for estimating vaccine take-up and other COVID-19-related outcomes (Bradley et al., 2021). However, numerous analyses have used HPS data for timely health and socioeconomic research (Berkowitz & Basu, 2021; C. S. Carpenter et al., 2022; Donnelly & Farina, 2021). Moreover, we utilize survey weights provided by the U.S. Census Bureau to estimate nationally representative results. Finally, the core of our analysis examines relative receipt of government assistance programs, comparing gay men, lesbian women, and bisexual individuals to their heterosexual counterparts during the COVID-19 pandemic, rather than establishing pre-COVID baseline estimates for economic status and government assistance utilization.

Our data are all self-reported, and thus, there may be selection effects associated with disclosing sexual orientation in an online survey. Secondly, our sample of respondents only includes non-institutionalized adults randomly selected for participation in an email survey among US households. This means that we are missing homeless adults; adults residing in institutionalized medical and incarceration facilities; and individuals without email addresses. Some sexual minority research has suggested that these exclusions may disproportionately affect LGB individuals since they report higher rates of homelessness (Corliss et al., 2011; Durso & Gates, 2012; Rosario et al., 2012), suggesting that our estimates on economic outcomes are likely a lower

bound for the disparities experienced by sexual minorities. Finally, for most of the programs assessed, we can only measure public assistance take-up through self-reported utilization rather than actual receipt and utilization, limiting the scope of our analysis. Some sexual minorities may receive community-based and charitable assistance through health, housing, and other service-based centers. More broadly, it is difficult to determine how much of our results can be attributed to the economic shocks of COVID-19 and their disproportionate impacts on sexual minorities or to preexisting economic disparities between heterosexual and sexual minority populations. We do control for recent job loss in Appendix Table 3 to proxy for pandemic-related economic disruptions and find that our main results stand (higher take-up of public assistance), suggesting that our results are not entirely driven by pandemic-related economic phenomena.

### 4.3 Methods

We begin our analysis with estimating descriptive statistics. Then, to estimate the relative economic status of sexual minorities compared to their heterosexual peers, we utilize a regression approach with multiple specifications, following other literature on LGBTQ economic position and access to public assistance (Badgett et al., 2013; Badgett, 2018; C. S. Carpenter et al., 2022). We specify estimation equation (1) in the following way:

$$y_i = \beta_0 + \beta_1 X_i + \beta_2 (GAY)_i + \beta_3 (BISEXUAL)_i + \beta_4 \gamma_s + \beta_5 \delta_t + \varepsilon_i \quad (1)$$

Where our outcome variables  $y_i$  are economic outcomes (log household income, employment status, FPL status, and subjective financial hardship) for individual  $i$  and  $X_i$  is a vector of individual characteristics. For all analyses, we stratify the sample by sex assigned at birth and control for indicator variables for gay men and lesbian women and bisexual individuals respectively. Due to this sex split, gay men and bisexual men are compared to heterosexual men while lesbian women and bisexual women are compared to heterosexual women. The coefficients of interest  $\beta_2$  and  $\beta_3$  represent the disparity in an economic outcome between gay/lesbian respondents and bisexual respondents, respectively, compared with heterosexual respondents.

Our preferred specification includes controls for the state of residence ( $\gamma_s$ ) and survey week ( $\delta_t$ ) to adjust for any state-specific or week-specific effects. Individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex assigned at birth, gender minority status, the total number of children in the household, urban-rural status, and education (four categories).

To estimate public assistance receipt and utilization, we estimated a Heckman selection model:

$$\begin{aligned} y_i &= \beta_0 + \beta_1 X_i + \beta_2 (GAY)_i + \beta_3 (BISEXUAL)_i + \beta_4 \gamma_s + \beta_5 \delta_t + \varepsilon_i \quad (2) \\ (FPL)_i &= \beta_0 + \beta_1 X_i + \beta_2 (GAY)_i + \beta_3 (BISEXUAL)_i + \varepsilon_i \quad (3) \end{aligned}$$

where our outcome variables  $y_i$  are various government assistance receipt and utilization outcomes for individual  $i$  and  $X_i$  is a vector of individual characteristics. We use the same individual level controls. However, we also adjust for selection into poverty to differentiate between differentials

based on needs or other factors. Hence, following Heckman selection methods, we estimate equation (3) using a probit model to obtain a correction term for this selection, and then adjust for this selection when estimating equation (2).

Finally, we restrict our sample to those below the federal poverty level and estimate equation (2) without the selection correction. In all regressions, we restrict our sample to complete cases. All regressions and descriptive statistics are weighted using HPS person weights computed by the U.S. Census Bureau in Stata version 16 (StataCorp, 2019), except for the Heckman models.

#### **4.4 Descriptive Statistics**

Table 1a presents descriptive statistics for the full sample by sexual orientation and sex. Table 1a shows that gay men and lesbian women are younger, have fewer children, and are more likely to be employed than their heterosexual peers. Table 1a also shows that bisexual women are younger than gay men, lesbian women and heterosexuals but are about as likely to have children as heterosexuals. Bisexual men also have lower incomes and lower likelihood of having children than heterosexuals. Additionally, bisexuals are more likely to have household incomes below the FPL than heterosexuals and gay/lesbian adults, especially bisexual women. These descriptive statistics display qualitatively similar patterns to those identified in other large population-based samples of LGB adults in the United States (Badgett et al., 2021).

Table 1b presents a demographic profile of self-identified LGB people with household incomes under the federal poverty level. In Panel A, we estimate that approximately 90.0% of men below the FPL identified as heterosexual, 5.7% identified as gay, and 4.4% identified as bisexual. Among this subsample, heterosexual respondents were the oldest (46.1 years), followed by gay men (42.4 years), and bisexual men (34.9 years). In Panel B, we estimate that approximately 88.7% of women below the FPL identified as heterosexual, 2.6% identified as gay or lesbian, and 8.7% identified as bisexual. Among women below the FPL, heterosexual respondents were the oldest (47.8 years), followed by lesbian women (37.3 years), and bisexual women (30.1 years). Heterosexual individuals experiencing poverty were more likely to be female than male (0.59 vs. 0.41). This pattern was even stronger among bisexual females and bisexual males (0.74 vs. 0.26). However, this sex breakdown is reversed for gay/lesbian adults experiencing poverty—they are more likely to be male than female (0.60 vs. 0.40). Motivated in part by this descriptive finding and prior literature that finds heterogeneous economic outcomes for men and women among sexual orientation identity groups, we stratify our analyses by sex.

Gay/lesbian adults experiencing poverty were also less likely to be Black, non-Hispanic than heterosexuals (0.14 vs. 0.19). Bisexuals experiencing poverty were less likely to be Black, non-Hispanic (0.11 vs. 0.19), and more likely to be non-Hispanic white (0.58 vs. 0.49) than heterosexuals. Marriage rates among those experiencing poverty were much lower for gay men and lesbian women (0.15) and bisexuals (0.17) than for heterosexuals (0.32). Additionally, even among those with incomes below the federal poverty level, educational attainment was higher for gay men, lesbian women, and bisexuals than for heterosexual participants.

## 5. Results

Below, we present a collage of evidence on the economic status and government assistance utilization of self-identified LGB people. In general, we only present coefficients that are statistically significant at the 5% level, and otherwise, coefficients will be noted as suggestive (if significant at the 10% level) or nonsignificant. We began above by examining the demographic and socioeconomic characteristics of LGB and heterosexual people of all household incomes and then restricted to those who are below the FPL. Next, we use multivariable regression to examine the relative economic status of LGB adults compared to heterosexual adults, while controlling for demographic characteristics. Our regression models compare the receipt and utilization of government assistance for LGB respondents with heterosexual respondents in a full sample controlling for demographic characteristics and a restricted sample of respondents who are below the FPL. We conclude the section by examining the receipt and utilization of the expanded child tax credit for LGB people compared to heterosexual people.

### 5.1 Economic Status of LGB Individuals

Table 2 presents our regression estimates of equation (1) for four economic outcomes. It asks the question of whether gay men, lesbian women, and bisexuals adults experience different outcomes from heterosexual adults after controlling for observable characteristics. The format of Table 2 is as follows: each column is a different outcome with the same regression model with indicators for sexual orientation, individual-level controls, as well as controls for state and survey week. Column (1) reports logged household income, calculated using interval regression; all other regressions were linear probability models. Column (2) reports employment; column (3) reports FPL status; and column (4) reports difficulty paying for expenses in the past week. We also provide the mean of each outcome below to contextualize the differences. The top panel estimates equation (1) for men, and the bottom panel displays the corresponding results for women.

The results in Table 2 indicate that bisexual men report 8.1 percent lower household incomes than otherwise comparable heterosexual respondents and that bisexual women report 8.9 percent lower household incomes than otherwise comparable heterosexual respondents. We also find suggestive evidence that lesbian women earn 3.2 percent less than otherwise comparable heterosexual women. In terms of poverty (column 3), we find that bisexual men are 2.2 percentage points more likely to report household incomes below the federal poverty level than heterosexual men. Relative to the mean of this outcome, poverty rates are 16 percent higher for bisexual men. Bisexual women are 2.9 percentage points more likely to report household income beneath the poverty line than heterosexual women (15 percent from the mean). Additionally, we find that bisexual men were approximately 6.1 percentage points more likely to report difficulty meeting expenses in the past week than otherwise comparable heterosexual respondents (24 percent relative to the mean); this differential was 6.0 percentage points for bisexual women (20 percent relative to mean). We also find a significant difference in self-reported financial hardship for gay men and lesbian women, despite not recording significant differences in other outcomes. Lesbian women were 4.6

percentage points more likely to report difficulty meeting expenses in the past week (15 percent relative to mean), while gay men were 3.4 percentage points more likely to do so (14 percent relative to mean). These results are broadly similar to what Chai & Maroto, 2020 find, although we find statistically significant differences for gay men, as well as lesbian and bisexual women.

In Appendix Table 4, we analyze differences in economic outcomes with gay men and lesbian women as the comparison group to formally test for heterogeneity between sexual minority subgroups. We find that bisexual men earn 7.8 percent less than gay men and that bisexual women earn 5.7 percent less than lesbian women, highlighting the unique economic adversity faced by bisexual sexual minorities.

## **5.2 Receipt and Utilization of Government Assistance**

We next describe the results of estimating equations (2) and (3), presented in Table 3. We estimate a two-step Heckman selection model, analyzing differences in receipt and utilization of public assistance programs while adjusting for selection into poverty status. This analysis examines whether gay men, lesbian women, and bisexual adults receive and use funds from public assistance programs at rates higher than those of otherwise comparable heterosexuals; all outcome regressions were linear probability models. The format of Table 2 is as follows: column (1) reports unemployment insurance utilization; column (2) reports stimulus check utilization; column (3) reports SNAP receipt; column (4) reports SNAP utilization; column (5) reports rental assistance utilization; and column (6) reports Medicaid coverage. We also provide the mean of each outcome below to contextualize the differences. The top panel estimates equation (2) for men, and the bottom panel displays the corresponding results for women. Additionally, within each panel we show the coefficients from the poverty selection model for sexual identity.

The results in Table 3 indicate that bisexual men (10.2 pp), bisexual women (5.4 pp), gay men (6.9 pp) and lesbian women (2.9 pp) are more likely to receive SNAP benefits than their comparable heterosexual peers. Similar results are found for SNAP utilization. We also find that gay men are 2.4 percentage points, bisexual men are 4.2 percentage points, and bisexual women are 1.9 percentage points more likely to utilize rental assistance than otherwise comparable heterosexuals. Finally, we find that gay men (6.9 pp) are more likely to be insured by Medicaid than otherwise comparable heterosexual adults.

We note that use of the Heckman selection model in this setting should proceed with caution—identification using functional form arguments relies on correct specification of the selection equation, often empirically implausible. While it would be preferable to use an instrument or exclusion restriction to weaken those assumptions, our data are limited, and so we are unable to use an observed variable that affects poverty status but not public assistance receipt/utilization to overcome this identification challenge. We note that recent literature has highlighted the possibility of weaker identifying assumptions without an exclusion restriction in the selection context (Escanciano et. al. 2016) and compare our corrected results with uncorrected results (reported in Appendix Table 6) to assess what we can learn from our analysis. Intuitively, because we observe

elevated poverty rates for bisexual men and women (Table 2), we would expect the uncorrected estimate to be larger for these populations, and the Heckman correction to attenuate the coefficients. We see this qualitative pattern for several outcomes, including unemployment insurance utilization, stimulus check receipt, and Medicaid. However, for other outcomes, we observe larger coefficients in corrected models, suggesting a role for misspecification or other unobserved variables. Thus, these results should be interpreted with caution, though the similar qualitative patterns of results across uncorrected and corrected results increases our confidence in the broader pattern of results.

In addition, we assess the sensitivity of the uncorrected public assistance results (reported in Appendix Table 6) to differences in unobservable variables. Recently, methods to bound the impact of unobservable variables have been developed and applied to a wide range of problems in applied microeconomics (Oster 2019; Altonji Elder Taber 2005). Typically, these methods consider coefficient and R2 stability as observable controls are added and use sensitivity parameters to assess the extent to which unobservable factors can account for the coefficient on a variable. In our context, we apply Oster (2019)’s estimator using Stata’s `reg sensitivity` package and consider what amount of variation in the outcome unobservable variables would need to explain (relative to observable variables) for the coefficient on a dummy variable for gay/lesbian status or bisexual status to be equal to 0.

Appendix Table 7 reports the breakdown point ( $\delta$ ) for these methods with each of our main public assistance outcomes. We pool results rather than stratify by sex to gain precision. Observable variables are age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, poverty status, and education (four categories), survey week, and state of residence. Unobservable variables would have to explain ( $\delta * 100$ )% of the variation that these observables do for the coefficient on Gay/Lesbian or Bisexual to be 0. We assume the maximum attainable R2 if controlling for both observables and unobservables to be 1, which is almost certainly an upper bound, given the possibility of measurement error. Thus, these results should be considered lower bounds of the true breakdown points.

For some of our results, the breakdown points do not seem especially high—unobservables would have to explain 1.4% of the variation that observables do for the gay/lesbian differential take-up of unemployment insurance utilization to be 0 (unsurprising, given the coefficients are near zero and nonsignificant). For other outcomes, however, unobservables would have to explain a significant portion of the variation that our rich set of observables do, and for SNAP and Medicaid, even exceed the observables threshold. Thus, we believe these methods suggest that even in a worst-case scenario of unobservable selection that is not captured by standard controls or the Heckman model, there is still evidence of differential take-up for LGB populations.

In Appendix Table 5, we analyze differences in public assistance take-up with gay men and lesbian women as the comparison group to formally test for heterogeneity between sexual minority

subgroups. Other than a 3.8 percentage point positive differential in Medicaid take-up for bisexual women, we do not find evidence of any consistent significant differences in public assistance take-up between gay men and bisexual men or lesbian women and bisexual women.

In Table 4, we examine regressions but restrict our sample to only those respondents who reported household incomes below the federal poverty level. We are estimating equation (1) using linear probability models. We find that gay men are 7.4 percentage points (22 percent relative to mean), and bisexuals are 7.2 percentage points (21 percent relative to mean) more likely to utilize SNAP than their heterosexual counterparts under the federal poverty line. Similar results are found for SNAP utilization, rental assistance, and Medicaid. We do not see comparable results for female sexual minorities below the federal poverty line—most estimates of differential utilization/receipt are close to zero and they are neither consistently positive nor negative.

### **5.3 Receipt and Utilization of the Child Tax Credit**

Table 5 displays the results of estimating equation (2) for those programs that involve children in the household or at school. It is important to note that we control for both the presence of children in the household using an indicator variable and the number of children in the household (a continuous variable), though our results are statistically similar and larger in magnitude when dropping these controls. The format of Table 5 is as follows: the left panel estimates equation (2) for the full sample, while the right panel estimates equation (1) for the sample restricted to those below the FPL. Columns (1) and (3) report receipt of the child tax credit as the outcome, while columns (2) and (4) display the utilization of a school lunch card as the outcome.

The results in Table 5 indicate that gay men are 1.6 percentage points less likely to receive the child tax credit (10 percent relative to the mean) than otherwise comparable heterosexual men, even while controlling for the presence of children and number of children in the household. A similar difference is not reported for bisexual men. We also find that lesbian women are 3.7 percentage points less likely to receive the child tax credit and that bisexual women are 2.1 percentage points less likely to receive the child tax credit. This result holds when restricting to the below FPL sample; we find lesbian women experiencing poverty are 8.3 percentage points less likely to receive the child tax credit (27 percent relative to the mean) than otherwise comparable heterosexual women. Additionally, we find that bisexual women experiencing poverty are 3.5 percentage points less likely to receive the child tax credit (12 percent relative to the mean) than otherwise comparable heterosexual women. In Appendix Table 1, we present results that are robust to restricting our sample to individuals who have a child present in the household; even among only adults with children in the household, gay men, lesbian women, and bisexual women are significantly less likely to receive the child tax credit than their heterosexual counterparts.

## **6. Discussion**

We used newly available data from a large, nationally representative sample of adults in the United States from the 2021-2022 U.S. Census Bureau's Household Pulse Survey to study the economic



outcomes and government assistance receipt and utilization of sexual minorities. Our regression models for economic outcomes, controlling for observable demographic characteristics, find that bisexual individuals have significantly lower household incomes, a higher likelihood of being below the federal poverty level, and higher self-reported financial hardship compared with otherwise similar heterosexual individuals. These results confirm previous literature that has found that bisexuals have worse economic outcomes than their heterosexual peers (Badgett et al., 2013, 2021).

Additionally, we find higher self-reported financial hardship for gay men and lesbian women, showing that the large literature on income differentials and economic status for gay/lesbian adults in the US also translates into higher levels of subjective economic distress (C. S. Carpenter, 2005; Chai & Maroto, 2020; Klawitter, 2015). We also report higher poverty rates for bisexuals, supporting prior research in this area (Schneebaum & Badgett, 2019; Uhrig, 2015). Our results improve on the economic status of sexual minorities literature by using large samples of self-identified LGB individuals (rather than using partnership status to infer sexual orientation), including a subjective measure of economic status (i.e., self-reported financial hardship and receipt/utilization of public assistance programs), and leveraging new and recent data collected during the COVID-19 pandemic.

We find that lesbian women earn approximately 3.2 percentage points less than otherwise observationally similar heterosexual women (though marginally significant), contradicting most research pre-2015, which had found wage premia for lesbian women (Drydakis, 2022; Klawitter, 2015). However, several more recent studies have found significant lesbian wage penalties (Bryson, 2017; Martell, 2019). Several others have highlighted this puzzling case of a disappearing or waning lesbian wage premium during a time of improving social acceptance and propose that reduced family support as a result of intolerance of their sexual identity may be hurting lesbian women (Drydakis & Zimmermann, 2020; M. E. Martell, 2019). Additionally, it is possible that earlier studies, which often relied on using same-sex couple status or cohabitation as a proxy for sexual orientation, may have suffered from selection problems if same-sex couples are positively selected on income relative to the broader LGB population, including single LGB adults (M. E. Martell, 2021). Because we use self-reported sexual identity, our results are less susceptible to this potential bias. Our findings confirm the empirical pattern of a lesbian wage penalty in a large, nationally representative survey of US adults during the COVID-19 pandemic and suggest that the emerging lesbian wage penalty may last.

Additionally, we find that LGB individuals utilize a range of government assistance programs at a higher rate than their heterosexual counterparts, even when allowing for selection into poverty status. These include SNAP and rental assistance for gay men, lesbian women, and bisexual individuals. This accords with prior work examining receipt of government assistance programs by sexual minorities, which has found higher rates of take-up for SNAP, TANF, and Medicaid (Badgett et al., 2013; Badgett, 2018; Brown et al., 2016). We add to the literature by evaluating several programs that have not, to our knowledge, been examined in the context of differential

take-up for sexual minorities, including unemployment insurance, stimulus payments, and rental assistance. We show that these results are robust to adjusting for selection into poverty status. Additionally, we demonstrate that when restricting to a low-income subsample, we continue to see differentials for gay and bisexual men, but not for lesbian and bisexual women. We find the largest differentials in rental assistance programs and unemployment insurance.

These generally higher rates of participation and utilization of government assistance programs among LGB people accord with the qualitative predictions we generate from the take-up literature. That the effects are more robust for gay and bisexual men suggests that social networks may play a significant role. As to which of the explanations we propose may explain the differentials, we cannot fully adjudicate, but there are several signs that some explanations are less plausible than others. For example, the take-up differentials are robust to controlling for educational attainment, suggesting that higher educational attainment is not the only mechanism that explains differential take-up of public assistance programs. Additionally, they are robust to controlling for recent job loss (see Appendix Table 3), suggesting that differential exposure to employment shocks from the COVID-19 pandemic do not fully explain the differentials we observe. This suggests that some combination of lessened stigma, fewer alternatives for financial support (low substitutability), and social network effects may be responsible for higher take-up of public assistance among sexual minorities.

We also find that gay men, lesbian women, and bisexual women are less likely to receive the advance child tax credit, an enhanced policy that has been promoted during the pandemic for its major impact on child poverty reduction (Corinth et al., 2021; Goldin & Michelmore, 2022). We found statistically significant disparities between sexual minorities and their heterosexual peers even when controlling for the presence of a child in the household, the total number of children in the household, or restricting to only households with children present. Additionally, the difference appears even among the subsample below the FPL. One potential explanation would be an asymmetry between how LGB individuals report a child in the household for the Household Pulse Survey and whether the Internal Revenue Service (IRS) recognizes that child for the eligibility determination used to establish recipients of the child tax credit. For example, if an LGB individual (or couple) was supporting an adolescent whose parents did not accept their sexual orientation or gender identity, they might report that child as “in their household” for the HPS, but the IRS would not recognize that relationship unless the child was legally adopted. These instances of informal family networks and “found family” are much more common among LGB individuals than among straight individuals (Jackson Levin et al., 2020; McConnell et al., 2015; Snapp et al., 2015). Unfortunately, we are unable to pinpoint specific mechanisms, but we encourage future studies to explore these important issues.

There are several limitations to our analysis. One is that we use self-reported survey data, which may lead to biased reporting of outcomes, especially of government assistance receipt (Celhay et al., 2022; Meyer et al., 2022). Due to a scarcity of ways to identify sexual minorities, the use of administrative data to confirm our findings is not currently possible, although as more surveys and

datasets collect sexual minority status, future work could use this data and administrative linkages to analyze if these public assistance differentials are also observed in administrative data. Additionally, our inability to fully adjudicate between the different explanations for the observed differentials is a major limitation. While we generate qualitative predictions guided by prior work in the program participation literature and examine which of these is supported by the HPS data, ideally, future work could instrument for each potential explanation using exogenous variation in the variable of interest (e.g., social networks) and examine whether disparities are explained by this exogenous variation. Finally, we use data from the COVID-19 pandemic, given that it is the only data source that allows us to study these differentials with sufficient power. However, as we discussed in Section 4.2, it is possible that the results we find are driven by the effects of the pandemic rather than pre-existing differences in outcomes. Future work can use data from other time periods to determine whether our results are observed outside the pandemic context.

We also derive several findings that may be of interest to policymakers. Estimating the take-up of public assistance programs is vital for modelling their microeconomic and macroeconomic impacts, as nearly all will have some degree of nonparticipation. Our analysis shows that one demographic group, LGB individuals, take up public assistance programs at higher rates than economically similar heterosexual individuals, and that this disparity is likely due to a combination of social network effects, lessened stigma, and increased reliance on public programs. Policymakers should consider how these factors were differentially established by the LGB community and whether there are policy or community actions (like utilizing social and peer networks to target public assistance) that may streamline public program participation for populations in greatest need. Additionally, policymakers may consider public policies that directly target LGB people (e.g., same-sex marriage) may consider our estimates of higher public program take-up; private benefits (e.g., health insurance) may crowd out public benefits when sexual minorities enjoy protections in employment, housing, and education. Finally, the disparities in receipt of the child tax credit for LGB people who have children point to an area where policymakers must examine whether current targeting and disbursement methods are not fully reaching LGB individuals who may be eligible for the program.

In addition, our limitations highlight the need for better data on LGBTQ populations and serve as a call to policymakers and administrators to incorporate instruments that measure sexual orientation and gender identity into survey and administrative data sources. Better data would allow for us to observe administrative rather than self-reported receipt of public assistance, offer a more complete picture of economic outcomes, and expand the set of data sources available to study the economic well-being of this population outside the context of the COVID-19 pandemic.

## **7. Conclusion**

Our findings confirm that bisexuals experience poorer economic outcomes and that gay men, lesbian women, and bisexuals take up government assistance programs at higher rates than their heterosexual counterparts in the U.S. context. Additionally, we find evidence that this pattern of

higher take-up among gay and bisexual men is maintained in a low-income subsample, not explained by selection into poverty, and is likely due to a combination of social network effects, lessened stigma, and increased reliance on public programs. Finally, we find that gay men, lesbian women, and bisexual women are less likely to receive the recently enhanced child tax credit than similarly situated heterosexuals. In so doing, our study adds to a growing literature on the experiences of LGB people in poverty. Our results further the call for more social science and policy research on the inequality and poverty experienced in sexual minority communities.

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**Table 1a: Descriptive Statistics, Household Pulse Survey, Stratified by Sexual Orientation and Sex**

Panel A: Full Sample, Men

	Heterosexual	Gay/Lesbian	Bisexual	P-Value
Sample Size	199224	11647	4723	
Weighted Percentage	92.5%	4.6%	2.9%	
Age	50.088	43.392	36.067	<0.001
Number of Children	0.630	0.196	0.414	<0.001
Race/Ethnicity				
White, non-Hispanic	0.669	0.652	0.676	
Black, non-Hispanic	0.093	0.079	0.061	
Asian, non-Hispanic	0.056	0.041	0.044	<0.001
All other races, non-Hispanic	0.031	0.037	0.044	
Hispanic	0.151	0.192	0.174	
Relationship Status				
Married	0.641	0.285	0.297	
Widowed, Divorced, or Separated	0.138	0.090	0.123	<0.001
Never Married	0.219	0.623	0.579	
Missing	0.003	0.002	0.001	
Child Present	0.344	0.116	0.241	<0.001
Gender Minority	0.001	0.040	0.075	<0.001
Urban Area	0.324	0.419	0.333	<0.001
Educational Attainment				
Less than High School	0.062	0.044	0.060	
High School Graduate	0.311	0.213	0.267	<0.001
Some College	0.292	0.348	0.407	
Bachelor's or higher	0.335	0.395	0.266	
Employment Status				
Employed	0.629	0.682	0.683	
Unemployed	0.098	0.126	0.124	<0.001
Not in Labor Force	0.261	0.182	0.189	
Missing	0.011	0.010	0.005	
Health Insurance				
Private Health Insurance	0.557	0.621	0.604	
Public Health Insurance	0.348	0.273	0.252	<0.001
Uninsured	0.090	0.099	0.141	
Missing	0.006	0.008	0.003	
Below FPL	0.134	0.168	0.207	<0.001

Source: Weeks 34-43 Household Pulse Survey, authors' calculations. Weighted means. Note average household income and poverty status are determined using the midpoint of each household income range or the lower limit of the range for those in the highest income range. P-values calculated using  $\chi^2$  tests for categorical variables and ANOVA tests for continuous variables.

Panel B: Full Sample, Women

	Heterosexual	Gay/Lesbian	Bisexual	P-Value
Sample Size	286275	7360	14567	
Weighted Percentage	91.5%	2.3%	6.2%	
Age	51.121	42.500	32.336	<0.001
Number of Children	0.700	0.472	0.665	<0.001
Race/Ethnicity				
White, non-Hispanic	0.657	0.667	0.661	
Black, non-Hispanic	0.119	0.106	0.074	
Asian, non-Hispanic	0.047	0.030	0.030	<0.001
All other races, non-Hispanic	0.036	0.050	0.069	
Hispanic	0.141	0.147	0.166	
Relationship Status				
Married	0.558	0.372	0.315	
Widowed, Divorced, or Separated	0.246	0.134	0.124	<0.001
Never Married	0.192	0.491	0.559	
Missing	0.003	0.004	0.002	
Child Present	0.380	0.259	0.369	<0.001
Gender Minority	0.001	0.059	0.048	<0.001
Urban Area	0.315	0.329	0.309	<0.001
Educational Attainment				
Less than High School	0.059	0.058	0.055	
High School Graduate	0.281	0.207	0.220	<0.001
Some College	0.305	0.338	0.397	
Bachelor's or higher	0.355	0.397	0.328	
Employment Status				
Employed	0.543	0.625	0.688	
Unemployed	0.108	0.120	0.121	<0.001
Not in Labor Force	0.338	0.242	0.190	
Missing	0.011	0.012	0.002	
Health Insurance				
Private Health Insurance	0.512	0.587	0.587	
Public Health Insurance	0.419	0.321	0.300	<0.001
Uninsured	0.064	0.090	0.111	
Missing	0.006	0.002	0.002	
Below FPL	0.183	0.209	0.264	<0.001

Source: Weeks 34-43 Household Pulse Survey, authors' calculations. Weighted means. Note average household income and poverty status are determined using the midpoint of each household income range or the lower limit of the range for those in the highest income range. P-values calculated using  $\chi^2$  tests for categorical variables and ANOVA tests for continuous variables.

**Table 1b: Descriptive Statistics, Household Pulse Survey, Stratified by Sexual Orientation and Sex**

Panel A: Below the federal poverty level (FPL), Men

	Heterosexual	Gay/Lesbian	Bisexual	P-Value
Sample Size	14203	1238	761	
Weighted Percentage	90.0%	5.7%	4.4%	
Age	46.098	42.365	34.881	<0.001
Number of Children	0.812	0.382	0.567	<0.001
Race/Ethnicity				
White, non-Hispanic	0.494	0.548	0.607	
Black, non-Hispanic	0.160	0.118	0.072	
Asian, non-Hispanic	0.051	0.038	0.008	<0.001
All other races, non-Hispanic	0.042	0.036	0.072	
Hispanic	0.254	0.260	0.242	
Relationship Status				
Married	0.355	0.125	0.190	
Widowed, Divorced, or Separated	0.235	0.119	0.175	<0.001
Never Married	0.406	0.755	0.630	
Missing	0.003	0.000	0.005	
Child Present	0.394	0.184	0.278	<0.001
Gender Minority	0.002	0.072	0.090	<0.001
Urban Area	0.320	0.385	0.325	0.084
Educational Attainment				
Less than High School	0.174	0.107	0.144	
High School Graduate	0.459	0.329	0.315	<0.001
Some College	0.259	0.374	0.392	
Bachelor's or higher	0.108	0.190	0.148	
Employment Status				
Employed	0.405	0.384	0.505	
Unemployed	0.253	0.310	0.243	<0.001
Not in Labor Force	0.329	0.297	0.251	
Missing	0.013	0.009	0.002	
Health Insurance				
Private Health Insurance	0.232	0.238	0.318	
Public Health Insurance	0.518	0.553	0.465	
Uninsured	0.238	0.188	0.211	0.068
Missing	0.012	0.022	0.006	

Source: Weeks 34-43 Household Pulse Survey, authors' calculations. Weighted means. Note average household income and poverty status are determined using the midpoint of each household income range or the lower limit of the range for those in the highest income range. P-values calculated using  $\chi^2$  tests for categorical variables and ANOVA tests for continuous variables.

Panel B: Below the federal poverty level (FPL), Women

	Heterosexual	Gay/Lesbian	Bisexual	P-Value
Sample Size	32068	976	2890	
Weighted Percentage	88.7%	2.6%	8.7%	
Age	47.764	37.333	30.732	<0.001
Number of Children	1.091	0.759	0.889	<0.001
Race/Ethnicity				
White, non-Hispanic	0.478	0.506	0.563	
Black, non-Hispanic	0.209	0.172	0.123	
Asian, non-Hispanic	0.028	0.017	0.030	<0.001
All other races, non-Hispanic	0.050	0.076	0.075	
Hispanic	0.235	0.229	0.209	
Relationship Status				
Married	0.288	0.178	0.159	
Widowed, Divorced, or Separated	0.378	0.166	0.157	<0.001
Never Married	0.331	0.651	0.684	
Missing	0.004	0.006	0.000	
Child Present	0.527	0.357	0.457	<0.001
Gender Minority	0.000	0.095	0.054	<0.001
Urban Area	0.301	0.299	0.264	0.094
Educational Attainment				
Less than High School	0.175	0.139	0.095	
High School Graduate	0.422	0.313	0.316	<0.001
Some College	0.301	0.388	0.447	
Bachelor's or higher	0.102	0.161	0.142	
Employment Status				
Employed	0.353	0.430	0.541	
Unemployed	0.218	0.224	0.178	<0.001
Not in Labor Force	0.414	0.336	0.281	
Missing	0.014	0.010	0.001	
Health Insurance				
Private Health Insurance	0.177	0.306	0.294	
Public Health Insurance	0.661	0.509	0.527	<0.001
Uninsured	0.150	0.181	0.176	
Missing	0.013	0.004	0.005	

Source: Weeks 34-43 Household Pulse Survey, authors' calculations. Weighted means. Note average household income and poverty status are determined using the midpoint of each household income range or the lower limit of the range for those in the highest income range. P-values calculated using  $\chi^2$  tests for categorical variables and ANOVA tests for continuous variables.

**Table 2: LGB Individuals, Especially Bisexuals, Experience Worse Economic Outcomes.**  
Full Sample

	(1) Log of household income	(2) Employed	(3) Below federal poverty level	(4) Financial hardship
<i>Male</i>				
Gay	-0.00297 (0.0170)	-0.00409 (0.00965)	-0.00140 (0.00862)	0.0344*** (0.0101)
Bisexual	-0.0812*** (0.0250)	-0.0172 (0.0144)	0.0216* (0.0124)	0.0613*** (0.0146)
Mean of Outcome	89810	0.633	0.138	0.251
N	215594	215594	215594	215594
<i>Female</i>				
Lesbian	-0.0322* (0.0183)	-0.00667 (0.0104)	0.00878 (0.00991)	0.0458*** (0.0108)
Bisexual	-0.0891*** (0.0146)	-0.00367 (0.00805)	0.0285*** (0.00786)	0.0603*** (0.00828)
Mean of Outcome	78744	0.554	0.188	0.308
N	308202	308202	308202	308202

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models, except for column (1), which uses interval regression. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .



**Table 3: LGB Individuals Access and Use Government Assistance at Higher Rates.**  
Full Sample, Heckman Selection Model

	(1) Unemployment insurance utilization	(2) Stimulus check utilization	(3) SNAP receipt	(4) SNAP utilization	(5) Rental assistance utilization	(6) Medicaid
<i>Male</i>						
Gay	-0.00347 (0.00712)	0.00846 (0.0127)	0.0692*** (0.0136)	0.0722*** (0.0114)	0.0244*** (0.00605)	0.0687*** (0.0151)
Bisexual	0.00999 (0.0182)	-0.0400 (0.0312)	0.102*** (0.0343)	0.0970*** (0.0292)	0.0418*** (0.0154)	0.0114 (0.0392)
<i>Below FPL</i>						
Gay	-0.00128 (0.0182)	-0.00128 (0.0182)	-0.000913 (0.0182)	-0.00128 (0.0182)	-0.00128 (0.0182)	0.00576 (0.0184)
Bisexual	0.218*** (0.0246)	0.218*** (0.0246)	0.217*** (0.0246)	0.218*** (0.0246)	0.218*** (0.0246)	0.224*** (0.0248)
Mean of Outcome	0.039	0.115	0.092	0.041	0.007	0.121
N	215594	215594	215594	215594	215594	215594
<i>Female</i>						
Lesbian	-0.0136* (0.00775)	-0.0102 (0.0130)	0.0293* (0.0158)	0.00915 (0.0149)	0.00670 (0.00819)	0.00276 (0.0157)
Bisexual	-0.00739 (0.00658)	-0.0169 (0.0110)	0.0535*** (0.0134)	0.0472*** (0.0126)	0.0191*** (0.00695)	0.00879 (0.0133)
<i>Below FPL</i>						
Lesbian	0.0638*** (0.0212)	0.0638*** (0.0212)	0.0656*** (0.0212)	0.0638*** (0.0212)	0.0638*** (0.0212)	0.0624*** (0.0215)
Bisexual	0.188*** (0.0142)	0.188*** (0.0142)	0.188*** (0.0142)	0.188*** (0.0142)	0.188*** (0.0142)	0.188*** (0.0143)
Mean of Outcome	0.041	0.121	0.148	0.087	0.013	0.194
N	308202	308202	308202	308202	308202	308202

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a Heckman two-step selection model. We present coefficient estimates and standard errors. The individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). We allow for selection into poverty status based on sexual identity, age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Table 4: Low-income LGB Individuals Access SNAP and Rent Assistance at Higher Rates.**  
Sample is limited to those respondents with household incomes below the FPL.

	(1) Unemployment insurance utilization	(2) Stimulus check utilization	(3) SNAP receipt	(4) SNAP utilization	(5) Rental assistance utilization	(6) Medicaid
<i>Male</i>						
Gay	-0.00975 (0.0168)	-0.0211 (0.0231)	0.0737*** (0.0274)	0.0703*** (0.0230)	0.0230** (0.0117)	0.109*** (0.0266)
Bisexual	0.0120 (0.0173)	-0.0129 (0.0256)	0.0715** (0.0332)	0.0910*** (0.0296)	0.0412** (0.0208)	0.0834** (0.0332)
Mean of Outcome	0.075	0.197	0.340	0.174	0.034	0.388
N	16202	16202	16202	16202	16202	16202
<i>Female</i>						
Lesbian	0.00814 (0.0199)	0.000320 (0.0284)	-0.00186 (0.0250)	-0.00544 (0.0230)	-0.00236 (0.0106)	-0.0424* (0.0257)
Bisexual	0.00184 (0.0131)	0.0165 (0.0177)	0.00267 (0.0169)	0.0126 (0.0156)	0.00716 (0.00740)	0.00688 (0.0175)
Mean of Outcome	0.077	0.213	0.468	0.303	0.050	0.536
N	35934	35934	35934	35934	35934	35934

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Table 5: Child-Related Programs are Differentially Accessed by LGB Populations.**  
Full Sample at left; FPL sample at right.

	Full Sample		<FPL Sample	
	(1) Child tax credit receipt	(2) School lunch card utilization	(3) Child tax credit receipt	(4) School lunch card utilization
<i>Male</i>				
Gay	-0.0161*** (0.00553)	0.000625 (0.00416)	-0.0199 (0.0160)	-0.0243* (0.0125)
Bisexual	0.00456 (0.00839)	0.0124 (0.00872)	-0.00203 (0.0221)	0.00451 (0.0232)
Mean of Outcome	0.161	0.032	0.191	0.101
N	215594	215594	14187	14187
<i>Female</i>				
Lesbian	-0.0369*** (0.00695)	0.000354 (0.00596)	-0.0827*** (0.0205)	-0.0139 (0.0193)
Bisexual	-0.0211*** (0.00567)	0.0139*** (0.00538)	-0.0350** (0.0151)	0.00119 (0.0143)
Mean of Outcome	0.233	0.061	0.302	0.161
N	308202	308202	31199	31199

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). Full sample regressions include a control for poverty status. We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

## Appendix

**Appendix Table 1: LGB Populations with Children Differentially  
Access Child-Related Programs.**

Full income sample with children at left; FPL sample with children at right.

	Full Sample		<FPL Sample	
	(1) Child tax credit receipt	(2) School lunch card utilization	(3) Child tax credit receipt	(4) School lunch card utilization
<i>Male</i>				
Gay	-0.106*** (0.0363)	-0.0265 (0.0185)	-0.130** (0.0554)	-0.123*** (0.0387)
Bisexual	0.0469 (0.0316)	0.000273 (0.0189)	0.00439 (0.0715)	-0.0725 (0.0461)
Mean of Outcome	0.558	0.073	0.410	0.181
N	51898	51898	3603	3603
<i>Female</i>				
Lesbian	-0.0968*** (0.0232)	0.00745 (0.0179)	-0.155*** (0.0538)	-0.0173 (0.0402)
Bisexual	-0.0301** (0.0135)	0.0221** (0.0109)	-0.0561* (0.0288)	0.0127 (0.0253)
Mean of Outcome	0.605	0.137	0.568	0.264
N	90610	90610	13380	13380

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). Full sample regressions include a control for poverty status. We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Appendix Table 2: Variable Creation**

Variable	Type	Coding
Gender Minority	Indicator	0) "Cisgender" 1) "Gender Minority"
Gay/Lesbian*	Indicator	0) "Heterosexual" 1) "Gay/Lesbian"
Bisexual*	Indicator	0) "Heterosexual" 1) "Bisexual"
Sex	Indicator	0) "Male" 1) "Female"
Age	Continuous	17-88
Race	Categorical	1) "White, Non-Hispanic" 2) "Black, Non-Hispanic" 3) "Asian, Non-Hispanic" 4) "All Other Races, Non-Hispanic" 5) "Hispanic"
Marital Status	Categorical	1) "Married" 2) "Widowed, Divorced, or Separated" 3) "Never Married" 4) "Missing"
Child <18 Present in Household	Indicator	0) "No Child in House" 1) "Child in House"
Educational Attainment	Categorical	1) "Less than High School" 2) "High School Graduate" 3) "Some College" 4) "Bachelor's or Higher"
Household Income	Range	1) "Less than \$24,999" 2) "\$25,000-\$34,999" 3) "\$35,000-\$49,999" 4) "\$50,000-\$74,999" 5) "\$75,000-\$99,999" 6) "\$100,000-\$149,999" 7) "\$150,000-199,999" 8) "200,000+" 9) "Missing"
Federal Poverty Line <sup>†</sup>	Indicator	0) "Above FPL" 1) "Below FPL"
Employment	Categorical	1) "Employed" 2) "Unemployed" 3) "Not in Labor Force" 4) "Missing"
Employed	Indicator	0) "Unemployed or Not in Labor Force" 1) "Employed"
Health Insurance	Categorical	1) "Private Health Insurance" 2) "Public Health Insurance" 3) "Uninsured" 4) "Missing"
Urban-Rural Status	Indicator	0) "Not in Metropolitan Statistical Area" 1) "In MSA"
State	Categorical	1) "Alabama" 2) "Alaska" 3) "Arizona" ....
Child Tax Credit Receipt	Indicator	0) "Did not receive Child Tax Credit" 1) "Received Child Tax Credit"
SNAP Receipt	Indicator	0) "Did not receive SNAP" 1) "Received SNAP"
Financial Hardship (Difficulty with Expenses)	Categorical	1) "Not at all difficult" 2) "A little difficult" 3) "Somewhat difficult" 4) "Very difficult"
Financial Hardship*	Indicator	0) "Not at all difficult or A little difficult" 1) "Somewhat difficult or Very difficult"

Unemployment Insurance Utilization <sup>‡</sup>	Indicator	0) "Did not utilize Unemployment Insurance" 1) "Utilized Unemployment Insurance"
Stimulus Payment Utilization <sup>‡</sup>	Indicator	0) "Did not utilize Stimulus Payments" 1) "Utilized Stimulus Payments"
Child Tax Credit Utilization <sup>‡</sup>	Indicator	0) "Did not utilize Child Tax Credit" 1) "Utilized Child Tax Credit"
School Meal Card Utilization <sup>‡</sup>	Indicator	0) "Did not utilize School Meal Card" 1) "Utilized School Meal Card"
Rental Assistance Utilization <sup>‡</sup>	Indicator	0) "Did not utilize Rental Assistance" 1) "Utilized Rental Assistance"
Medicaid Utilization	Indicator	0) "Did not report Medicaid as a source of health insurance" 1) "Report Medicaid as a source of health insurance "

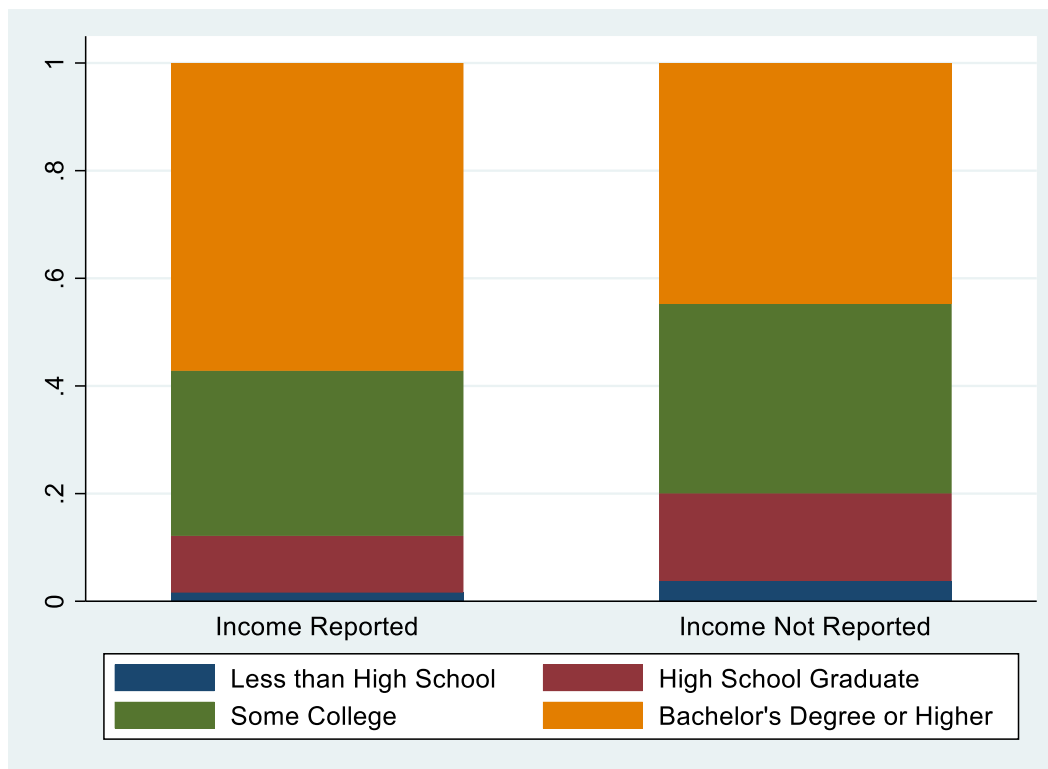
Notes: \*Bisexuals are coded as missing (.) for the gay/lesbian indicator variable, and vice versa so that comparisons are to a strictly heterosexual group. <sup>†</sup>Following C. S. Carpenter et al., 2022, we use the midpoint of the reported income range as the value for that respondent, and then compare that to their federal poverty threshold (determined by household size and number of children in the household). If participants do not exceed the threshold, they are coded as below the federal poverty line. <sup>\*</sup>We dichotomize the financial hardship variable into an indicator variable for ease of interpretation and following Garner et al., 2020b; Kim, 2021. <sup>‡</sup>Respondents were asked which of the following sources they and their household members used to meet their spending needs. They could choose multiple options including unemployment insurance (UI), stimulus payments, SNAP, school meal cards, and governmental rental assistance. If they selected one of these options, they were coded as having utilized that government assistance program.

**Appendix Table 3: LGB Public Assistance Differentials, Controlling for Recent Job Loss**

	(1) Unemployment insurance utilization	(2) Stimulus check utilization	(3) SNAP receipt	(4) SNAP utilization	(5) Rental assistance utilization	(6) Medicaid
<i>Male</i>						
Gay	0.00576 (0.00533)	0.00134 (0.00779)	0.0267*** (0.00697)	0.0213*** (0.00522)	0.00555** (0.00230)	0.0333*** (0.00742)
Bisexual	0.0133 (0.00861)	0.0197* (0.0112)	0.0380*** (0.0118)	0.0371*** (0.00922)	0.0137*** (0.00525)	0.0322*** (0.0108)
Mean of Outcome	0.039	0.115	0.092	0.041	0.007	0.121
N	215594	215594	215594	215594	215594	215594
<i>Female</i>						
Gay/lesbian	-0.00235 (0.00589)	0.0208** (0.00958)	0.0178** (0.00824)	0.00726 (0.00684)	0.00128 (0.00296)	-0.00159 (0.00999)
Bisexual	0.00248 (0.00443)	0.0159** (0.00698)	0.0214*** (0.00665)	0.0188*** (0.00560)	0.00224 (0.00217)	0.0351*** (0.00752)
Mean of Outcome	0.041	0.121	0.148	0.087	0.013	0.194
N	308202	308202	308202	308202	308202	308202

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Coefficients are from linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include recent job loss age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, poverty status, and education (four categories). We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Appendix Figure 1: Educational Attainment, by Income Reporting**



Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). The first bar reports the distribution of educational attainment for the sample that reported their income (our primary analytic sample), and the second reports that distribution for the sample that did not report their income.



**Appendix Table 4: Bisexuals have Lower Incomes than Gay men and Lesbians.**

Full income sample with children at left; FPL sample with children at right.

	(1) Log of household income	(2) Employed	(3) Below federal poverty level	(4) Financial hardship
<i>Male</i>				
Heterosexual	0.00297 (0.0170)	0.00409 (0.00965)	0.00140 (0.00862)	-0.0344*** (0.0101)
Bisexual	-0.0782*** (0.0293)	-0.0131 (0.0166)	0.0230 (0.0144)	0.0269 (0.0171)
Mean of Outcome	89810	0.633	0.138	0.251
N	215594	215594	215594	215594
<i>Female</i>				
Heterosexual	0.0322* (0.0183)	0.00667 (0.0104)	-0.00878 (0.00991)	-0.0458*** (0.0108)
Bisexual	-0.0569** (0.0231)	0.00299 (0.0127)	0.0198 (0.0124)	0.0145 (0.0132)
Mean of Outcome	78744	0.554	0.188	0.308
N	308202	308202	308202	308202

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, sex, urban-rural status, gender minority status, the total number of children in the household, and education (four categories). We also control for survey week and state of residence. \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Appendix Table 5: Sexual Minority Subgroups have Similar Public Assistance Takeup**  
Full Sample

	(1) Unemployment insurance utilization	(2) Stimulus check utilization	(3) SNAP receipt	(4) SNAP utilization	(5) Rental assistance utilization	(6) Medicaid
<i>Male</i>						
Heterosexual	-0.00708 (0.00551)	-0.00193 (0.00783)	-0.0272*** (0.00693)	-0.0215*** (0.00520)	-0.00558** (0.00230)	-0.0338*** (0.00743)
Bisexual	0.00911 (0.00993)	0.0192 (0.0131)	0.0121 (0.0132)	0.0162 (0.0102)	0.00823 (0.00560)	0.000260 (0.0125)
Mean of Outcome	0.039	0.115	0.092	0.041	0.007	0.121
N	215594	215594	215594	215594	215594	215594
<i>Female</i>						
Heterosexual	0.000886 (0.00605)	-0.0213** (0.00962)	-0.0186** (0.00818)	-0.00771 (0.00680)	-0.00141 (0.00296)	0.000759 (0.00990)
Bisexual	0.00746 (0.00763)	-0.00372 (0.0117)	0.00524 (0.0102)	0.0127 (0.00852)	0.00119 (0.00352)	0.0381*** (0.0120)
Mean of Outcome	0.041	0.121	0.148	0.087	0.013	0.194
N	308202	308202	308202	308202	308202	308202

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, poverty status, and education (four categories). We also control for survey week and state of residence.

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Appendix Table 6: LGB Individuals Access and Use Government Assistance at Higher Rates.**  
Full Sample

	(1) Unemployment insurance utilization	(2) Stimulus check utilization	(3) SNAP receipt	(4) SNAP utilization	(5) Rental assistance utilization	(6) Medicaid
<i>Male</i>						
Gay	0.00708 (0.00551)	0.00193 (0.00783)	0.0272*** (0.00693)	0.0215*** (0.00520)	0.00558** (0.00230)	0.0338*** (0.00743)
Bisexual	0.0162* (0.00889)	0.0211* (0.0112)	0.0393*** (0.0119)	0.0377*** (0.00920)	0.0138*** (0.00523)	0.0341*** (0.0107)
Mean of Outcome	0.039	0.115	0.092	0.041	0.007	0.121
N	215594	215594	215594	215594	215594	215594
<i>Female</i>						
Lesbian	-0.000886 (0.00605)	0.0213** (0.00962)	0.0186** (0.00818)	0.00771 (0.00680)	0.00141 (0.00296)	-0.000759 (0.00990)
Bisexual	0.00658 (0.00448)	0.0175** (0.00697)	0.0238*** (0.00665)	0.0204*** (0.00560)	0.00260 (0.00216)	0.0374*** (0.00752)
Mean of Outcome	0.041	0.121	0.148	0.087	0.013	0.194
N	308202	308202	308202	308202	308202	308202

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each column is from a separate regression using the HPS person weights. Linear probability models. We present coefficient estimates and standard errors, robust to heteroskedasticity. The individual-level controls include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, poverty status, and education (four categories). We also control for survey week and state of residence.  
\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ .

**Appendix Table 7: Breakdown Points for Selection on Unobservables**  
Full Sample

	(1) Unemployment insurance utilization	(2) Stimulus check utilization	(3) SNAP receipt	(4) SNAP utilization	(5) Rental assistance utilization	(6) Medicaid
Gay/Lesbian N	0.014 504,506	0.226 504,506	29.889 504,102	0.998 504,506	0.064 504,506	1.410 460,091
Bisexual N	0.025 504,789	0.023 504,789	0.162 500,438	0.116 504,789	0.031 504,789	0.188 460,953

Notes: Data are from the Weeks 34-43 Household Pulse Survey (HPS). Each entry applies the Oster (2019) method to bound selection on unobservable variables and reports the breakdown point ( $\delta$ ). The individual-level controls used as observables include age, age squared, race, ethnicity, relationship status, the presence of a child in the household, sex, urban-rural status, gender minority status, the total number of children in the household, poverty status, and education (four categories). We also control for survey week and state of residence.