Local Political Context and Pro-Palestinian university

encampments



By Carter Castillo

A Senior Honors Thesis Submitted to the Department of Political Science, University of California, San Diego

March 31, 2025

Acknowledgments

I want to start out thanking my advisor and mentor Dr. Karen Ferree. This project would not exist without your help and support. When I was a doe-eyed anxious freshman, you saw potential in me, and you encouraged me to think about an honors thesis. Your constant guidance and advice, and the level of attention and support you gave to me was not only invaluable to the creation of this project, but my overall experience as a student at UCSD. You are such an incredible mentor, and I remember coming back from every single meeting we had together about my honors thesis and the first thing I would do is think of how lucky I am to learn from someone who is not only incredibly intelligent and experienced in her field, but someone who I could tell really cared about my own personal growth as a student of the political sciences. From the bottom of my heart, thank you.

To the seminar advisors, Dr. LaGina Gause and Dr. James Fowler, I want to thank you for helping create a comfortable atmosphere where we students in the seminar could share our research and give constructive feedback. I also want to thank Dr. Gause and also Dr. David Fisk for helping point me in the right direction with forming the foundations of my research. Without those solid foundations, this project would not have been possible.

Finally, I want to thank my family. To my mom and dad, who tolerated my frantic ramblings about my honors thesis for months, thank you for tolerating me. To my Grandma, I know this year has been difficult, but I want you to know that I would not be able to be here without the constant love and support you gave me throughout my entire life. Your support was crucial to allowing me to complete this project. In the midst of my most stressful moments, I always knew I could reach out to my grandma.

Local Political Context and Pro-Palestinian university

encampments

Acknowledgments	2
1 Introduction	5
2 Literature Review	7
2.1 Introduction	7
2.2 Resource Mobilization and Political Process Theories	7
2.3 Political Environment and Opportunities	10
2.4 Leftism and Protests	12
2.5 University Protests	13
2.6 Importance of Mixed Methods Research Design	14
3 Hypotheses	15
4 Research Design	15
4.1 National Data Set	16
4.1.2 Independent variable	17
4.1.3 Dependent variable	18
4.1.4 Control Variables	19
4.1.5 Method of Analysis	19
4.2 Only University Encampments	20
4.2.1 Independent Variable	20
4.2.2 Dependent Variable	20
4.2.3 Control Variables	21
4.2.3 Method of Analysis	22
4.3 Qualitative Data	22
5 Data Analysis	24
5.1.1 National Data Frequencies	24
5.1.2 National Dataset Hierarchical Regression	26
5.2.1 Encampment Only Hierarchical Regression	
5.2.3 Analysis of Force Variable	32
5.3 Qualitative Data	34
6 Conclusion	38
6.1 Discussion	38
6.2 Limitations	40
6.3 Future Research	41
Appendices	44
References	60

Table of Contents

<u>1 Introduction</u>

The Pro-Palestine encampments in Spring of 2024 were one of the largest student-centric protests of the 21st century so far, with estimates of thousands of students across the country participating in protest encampments, and thousands were arrested (Habeshian 2024). The protests were clearly political, with many encampments focusing their ire on the current Democratic administration's foreign policy regarding Israel and Palestine (Popli 2024). However, not every university had an encampment, and not every encampment had the same level of participation.

Did the political context of a university impact whether its students decided to have an encampment at that university? Existing evidence is mixed, with previous research finding varying levels of support of the impact environmental political factors have on protests (Edwards 2014, 90; Sabine 2006). Additionally, due to the recency of these protests, there is not much literature regarding the pro-Palestine movement's recent student encampments.

I argue that local political context matters in whether an encampment occurred at a university campus. More specifically, that the more Democratic an area, the more likely there will be an encampment that will be larger. This is because broadly, the current Democratic party is more open to protest, seen with large portions of its base supporting the Black Lives Matter Movement, and recent polling data that shows that Democrats are more sympathetic towards Palestinians than Republicans (PEW Research Center 2018; Gallup 2023). This leads to two possible causal mechanisms that I label: the preference mechanism and the punishment mechanism. The preference mechanism claims that greater amounts of partisan allies in an area means there are more potential people that would start or join a protest. Meanwhile, the punishment mechanism asserts that in environments where the partisan makeup of local and regional governments are opposed to certain social movements, those social movements are less likely to generate protests because of the fear of repression by potential protestors.

In order to test my argument, I utilized a mixed methods research design, utilizing both quantitative and qualitative methods to help test my argument. For the quantitative analysis, I ran a hierarchical regression, with data collected from encampments from over 1,200 4-year universities in the U.S. compared with the university's 2020 county voting record. For the qualitative data, interviews were conducted with members of the encampment at UCSD, to identify common patterns in themes and identify true process and motivation.

The quantitative data supports my hypothesis that democratic political contexts made encampments more likely, and also larger. The results provide some evidence that suggests that political context can impact whether social protests occur, especially when the partisan political context is one that is more open towards protest.

This research sits within a gap in the political science field regarding the student-led Pro-Palestine movement. It also works to contribute knowledge of social movements in the specific field of political science. Future research could try to interview encampment participants from other universities, particularly those from red states where the encampments were forcefully ended rapidly, such as in Georgia and Texas. Additionally, more research could be done in understanding the outcomes of the encampments, namely whether they ended in a negotiated agreement or ended by force.

2 Literature Review

2.1 Introduction

There is extensive literature examining social movements, protests, and contentious politics. This literature review will focus on the relationship between political science and social movement literature. It will then analyze the specific theories of social movements as they relate to political opportunities and partisan electoral politics. Third, this review will analyze ideological political beliefs in relation to how likely an individual is to join a protest. Then, this review will look at University demonstrations specifically. Finally, this literature review will discuss the importance of mixed methods design in analyzing social movements from a political science perspective.

This paper covers contemporary events that are extremely contentious. In an attempt to remain as unbiased as possible, the term pro-Palestine will be used to refer to those who participated in or heavily sympathize with the encampments that occurred at university campuses worldwide. While this term does not fully encapsulate a lot of the nuances and even disagreements within the pro-Palestine movement, as Chenoweth et al. in 2024 notes that alternative terms are "even less satisfying characterizations." For example, Chenoweth et al. finds that generalized categorizations of the pro-Palestinian movement as "anti-Israel" are "empirically incorrect."

2.2 Resource Mobilization and Political Process Theories

The field of contentious politics, which includes all forms of protest, has been interdisciplinary in nature, with overlap between sociology, political science, anthropology, history, and even social psychology (Tarrow 2021). However, certain research of contentious politics is uniquely suited to the field of political science. In her journal article Kateřina Vráblíková writes that sociologists use political science concepts for analysis of social movements, which gives a major boon to conducting research on social movements within the political science field (Vráblíková 2017, 5-6).

Historically, there are three analytical approaches to studying social movements: class analysis, role theory, and structural functionalism (Walder 2009, 394-395). What all three approaches share in common is that they try to relate social structures to the character of the social movements; however, they all failed to accurately predict social movements, and as a result, resource mobilization theory became the predominant theory (Walder 2009, 396).

Resource mobilization has its roots in the rational choice theory of the economist Mancur Olson. Traditional theories of why protests occurred relied on the idea that people had grievances and deprivation from their government that naturally spurred them to protest in response. However, Olson challenged this traditional paradigm with his introduction of rational choice theory. Rational choice theory focuses on the fact that the members of a social movement are rational actors, and they have to weigh material costs and benefits when deciding whether to join a protest (Mueller 1992, 3). In fact, this idea that individual activists are rational actors was recently corroborated in a study of the 2020 Black Lives Matter protests by Chenoweth et al. published in 2022. This study found that protestors made a deliberate decision based on potential costs and benefits associated with participating in the protests (Chenoweth et al. 2022, 21 and 27).

Rational choice theory laid the foundations for resource mobilization theory, which focuses on the ability for a movement to gain participation through material and immaterial resources. As a result, many analysts look for factors outside of the social movement that could potentially inhibit or enhance the potential for a social movement to mobilize (Meyer 2004). More specifically, resource mobilization theory focuses on what resources are available for a movement, how the movement organizes, how the state facilitates or impedes mobilization, and what the outcomes of the protest/social movement are (Mueller 1992, 3-4). For the scope of this paper, the predominant focus will be in determining if the state/local political environment impeded or aided mobilization for these protests, with some attention drawn to the outcomes of the protests as well.

Political process theory is considered an extension of resource mobilization theory, as it not only centers the mobilizing problem as central, but it also assumes there is an internal cost benefit analysis that rational actors make when deciding whether to participate in a form of social movement (Edwards 2014, 79-80). However, political process differs from resource mobilization theory as instead of focusing on internal resources a movement has, political process theory analyzes the political context in relation to protests and social movements (Edwards 2014, 79-80). Moreover, proponents of political process theory assert that the strategies and decisions employed by activists do not occur in a vacuum, and thus that the political context matters in the mobilization and outcome of potential protest (Meyer 2004). Essentially, political process theory asserts that without a favorable political context, then protests will struggle to achieve any desired outcomes. This could be for two reasons. First is the system level explanation, wherein politicians and bureaucrats implement policies that are favorable or repressive towards social movements because of the community support in an area for that social movement. This is what I am calling the punishment mechanism. A non mutually exclusive alternative is that of the community support itself, where more support for perceived partisan allies means that protests have more resources and people willing to join their own movement. This is what I am calling the preference mechanism. While this study does

empirically attempt to isolate these two causal mechanisms, they can both be measured by viewing community support for partisan political candidates.

2.3 Political Environment and Opportunities

This paper focuses on analyzing the local political environment of US universities. Political environment and political context are terms in the literature that are generic, with researchers tending to avoid large conceptual definitions in favor of identifying specific variables needed for their specific endeavor (Eisinger 1973, 11-12; Meyer 2004). Nevertheless, the idea of studying a political environment is important, as Eisinger notes that environments can constrain activity and thus deter mobilization (Eisinger 1973, 11-12). Essentially, protests do not occur in a vacuum, and while definitions may vary depending on the specific research question, it is nonetheless important to analyze the role of partisan politics and state actors in mobilization of protests. In fact, previous research has operationalized the political environment to mean the support presidential candidates received within a local area (Huckfeldt 1995, 1026). This legitimizes this studies operationalizing of the political environment on the national scale by focusing on county level data for the 2020 presidential election.

One of the major ways in which political science can uniquely answer puzzles in regards to resource mobilization is by analyzing political opportunities and structures. Traditionally, researchers in political science have defined political opportunities to reflect the 'openness' or 'closedness' of state institutions (Kitschelt 1986, 61). The openness and closedness of an institution often refers to the willingness of governmental forces to crackdown on protest, and the willingness of legislatures to work with activists. It is thus argued that the more open an institution, the more conducive it is to demonstrations and protests (Kitschelt 1986, 61-62). This is because opportunities, policies, and environments that are conducive to protest in political

structures shapes the orientations, growth, and success of movements, as seen empirically during the civil rights movement in 1950s and 1960s (Walder 2009, 403). Moreover, it is hypothesized that the state can deter mobilization through threats of violence, however the research is mixed in support of this hypothesis (Sabine 2006, 1). This only adds to the potential knowledge this research could contribute.

There is also debate in current social movement literature regarding whether there is enough attention paid to electoral party politics and political context (Vráblíková 2017, 17). This further allows this thesis to sit within the context of this larger debate and hopefully provide evidence that demonstrates a relationship between politics and social movements. This is especially true considering that only a few studies have been conducted examining the effect that partisanship has on protests (Silver 2023).

Evidence from the Black Lives Matter Protests in 2020 indicate that partisanship influenced support of police repression (Silver 2023). This empirically suggests that there is a possible relationship between partisanship and the responses towards protests dependent on how the protest itself is viewed. This is potentially because the public support and response signals to policymakers what methods and posture towards the protest are politically viable for them to use (Silver 2023). This suggests that if a local political context is strongly opposed to a protest and its aims, then their attitude towards suppression and repression by governmental entities would be more favorable. This would suggest that protests that occur in local contexts with a larger base of partisan opposition would yield fewer protests, and that the ones that materialize would also be smaller. Likewise, in areas with more numerous political allies there would be more and larger protests. However, these ideas often lack rigorous empirical backing due to a lack of current available literature on the relationship between partisanship and protest existence and size. This research thus seeks to understand the relationship between partisanship and whether a protest materializes in the context of the 2024 Spring Pro-Palestine university encampments.

2.4 Leftism and Protests

The left-wing, defined as a broad range of political ideologies that seek to mitigate inequality (occasionally with ties to Marxism and or socialism) is historically associated with a higher propensity to mobilize in favor of protest (Torcal et al. 2016). Moreover, empirical evidence suggests that individuals protest under more right-wing governments than under left-wing governments (Torcal et al. 2016). This is because of two possible reasons, which may not inherently be conflictual. One is that the historical legacy of leftism values protest as this was a common theory of power employed by leftists (Kostelka 2019, 1680). The second explanation is that the ideological objectives of the left naturally align themselves with forms of contentious politics such as protest (Kostelka 2019, 1680-1681).

This situates this research well in trying to understand more local political contexts to determine if this finding remains true with right-wing local political contexts. It thus can be argued that the more left-wing a political environment, the more resources are available to protests, namely in the form of actual participants. This is especially true are there is a growing increase in polarization between the two major political parties in the U.S., with some members of the center-left Democratic party aligning themselves more with the pro-Palestine movement (Rynhold 2020). This occurrence lends further credence to the idea that the more votes the Democratic candidate received, the more successful the mobilization of a pro-Palestine encampment.

In the United States, between the two major parties the Democratic party is considered the party to be more in line with ideological values of the left, with policy proposals typically more centered on some level of economic redistribution and reducing inequality in comparison to Republicans' more laissez faire approach (Zacher 2024). While generalizing, this does put the Democratic party closer as ideological allies to leftism and progressivism than the Republican party. Additionally, contemporary polling data seems to strongly indicate that Democrats tend to be more sympathetic towards Palestine and more critical towards Israel than Republicans (PEW Research 2018; Gallup 2023). This is why I argue that Democrats are more sympathetic and even supportive of the Pro-Palestine Encampments than Republicans, which leads into the hypothesis.

While partisan voting relationships may not perfectly overlap with political ideology, this paper seeks to rectify that throughout its design by utilizing a mixed methods approach and using two cuts of quantitative data.

2.5 University Protests

There is much evidence that supports that tertiary education drives political activism and collective action, making the University a unique environment to analyze the effect of political structures and opportunities on protests (Dahlum and Wig 2021). Dahlum and Wig specifically identify social networks, organizations, opportunity costs, and focal points as factors that university's influence in creating a more conducive environment for protest, yet they also note that the link between tertiary education and mass protest is still poorly understood due to a lack of large-N studies and a focus on political membership operationalized as party membership. This leads to a gap in the literature that this paper can sit within by analyzing the recent pro-Palestine protests that occurred in Spring of 2024.

In trying to understand the size of Pro-Palestine demonstrations, Chenoweth et al. hypothesized that the actual actions of Israel and the U.S. in Gaza lead to the wide-scale pro-Palestine mobilization. However, this ignores the more pragmatic local material political context that impacts not only if a pro-Palestine demonstration occurs, but also its ability to grow in size, which is what this paper seeks to examine. Chenoweth et al. also recognizes that while the terms pro-Israel and pro-Palestine may flatten nuances within both movements, they are still the most accurate terms available.

While historically there is limited research from political science analyzing University protests in The United States, McCarthy et al. analyzed numerous University anti-war protests and demonstrations during the Vietnam war. They identified protest size as the key independent variable in analyzing when police crack down on protests (McCarthy et al. 2007, 278). While the scope of this paper is different, McCarthy et al.'s study does protest size as a relevant variable in studying University protests and also gives possible clues as to why repressive political environments would try and intentionally mitigate the growth of a University protest.

2.6 Importance of Mixed Methods Research Design

As Vráblíková identifies, the lack of quantitative and qualitative mixed methods research has limited the capacity of political science to empirically answer questions as it pertains to social protest (Vráblíková 2017, 24). While the structure of social relations is best studied with quantitative methods, qualitative methods are better to understand actions of individual agents and motivations (Thaler 2017, 60). This lends itself to the scope of this paper, which seeks to understand the relationship between protest size and political context. Quantitatively, this paper will look at a large-N sample and quantitatively analyze relationships between protest size and operationalized quantitative data for political context, while the qualitative side can be used to further isolate causality with political context in directly interacting with organizers and participants of the pro-Palestine demonstrations. This method combines the strengths identified in the literature as noted by Vráblíková and Thaler.

<u>3 Hypotheses</u>

There is one broad overarching hypothesis for this paper, which is as follows: A more partisan democratic political context results in more and larger Pro-Palestine encampments. This overarching hypothesis is consistent with at least two possible mechanisms that I forward. The first I label the preference mechanism, which is that a more democratic context means more democratic voters, whose preferences are more likely to be in support of the pro-Palestine movement and thus start and join in the pro-Palestine encampments. A second mechanism is what I call the punishment mechanism, which asserts that a more democratic context means there is less fear of punishment and repression towards protestors, alleviating significant costs in their decision to join a pro-Palestine encampment.

For the national level data set, the hypothesis is as follows: The universities in counties that voted for Joe Biden in states that voted for Joe Biden in 2020 are more likely to have a Pro-Palestine encampment.

I then hypothesize that among just the universities that had an encampment, those located in counties with a higher vote share for Joe Biden in the 2020 election will have a greater number of participants in said encampments.

The goal of the qualitative data is to get further insight on which causal mechanisms are more at play, and to contextualize the quantitative data. This goal ties back into the broader hypothesis that left-wing politics were more conducive to mobilization of the pro-Palestine encampments.

4 Research Design

The research question for this paper is: How does the local political context of a given university impact the size of a pro-Palestine encampment at the same university? I broadly hypothesize that the more left-wing the local political context, the larger the pro-Palestine encampment. More specifically, I hypothesize that this political context extends to both the state and county level. I thus hypothesize that the greater the vote percentage in a county the more likely a university will have a Pro-Palestine encampment. I also hypothesize that states that are controlled by Democratic governors are also more likely to have an encampment. Additionally, I hypothesize that the greater the vote for Joe Biden the more participants in the Pro-Palestine encampment. As explained in the literature review, there are a few potential causal mechanisms that justify these hypotheses. The first is that being in an environment with more supporters means there are more resources and planning available to execute a protest and more desire to protest this cause in the first place. The second is that being an environment where people are largely supportive or apathetic to the Pro-Palestine protest means that there will be less coercive pressure from governmental and institutional forces that would otherwise work to suppress or mitigate the existence and size of a Pro-Palestine encampment. This punishment mechanism also relies on the state level factors, specifically control over the state government, which connects back to my hypothesis regarding states whose governors are Democrats.

In order to best answer the research question and eliminate possible confounding variables, this research is conducted across three levels in a mixed methods design. Each level or cut of data will become more specifically targeted in its selection in the essence of an inverted pyramid, starting with the most broad large-N analysis and moving into a smaller-N sample with more targeted data collection, and finally ending with a case study with qualitative data.

4.1 National Data Set

The first and broadest cut involves universal data for nearly every single University in the United States. University is defined according to the U.S. Census Bureau is a university engaged in "furnishing academic courses and granting degrees at baccalaureate or graduate levels," ("North American Industry Classification System"). Additionally, I only included universities that had a recorded population of at least 1,000 people to sharpen the focus of the dataset. This dataset seeks to answer the specific hypothesis that: The greater the average vote percentage that the Democratic candidate for president received in the counties the University resides in, the greater the number of participants in the university's Pro-Palestine encampment.

I used an exhaustive list of every university according to the North American Industry Classification System (NAICS) from Opensoft. Only schools who had a physical location were selected, as the scope of the research question and hypotheses hinges upon the university occupying a physical space. This gave me a dataset with an N value of 1424.

4.1.2 Independent variable

The independent variable is the local political context, which in this case is operationalized as the partisan voting record in the county that encompasses the university campus in 2020. For this set of data, the local political context is operationalized to refer to the partisan presidential voting record in 2020, where votes for Democratic candidate Joe Biden are understood as liberal/left-wing, while votes for Republican candidate Donald Trump are understood as conservative/right wing. The terms "Blue" and "Red" are operationalized in this paper as shorthand for places that voted more for Biden or Trump respectively.

The raw vote totals for Joe Biden, Donald Trump, and other candidates were recorded for each county in which a university resides based upon each state's secretary of state office. This was then turned into percentages, yielding the independent variable I labeled Biden Vote percentage, referring to the percent Biden received in a given county. I used the county listed by the NAICS dataset. I then used this raw vote total to create dummy variables. Additionally, counties were coded as 0 if Trump received more votes and as a 1 if Biden received more votes. Once this was done, I then coded each university based on whether it was in a state where Trump received at least a plurality of the vote (coded as a 0) or in a state where Biden received at least a plurality of the vote (coded as a 1). I similarly coded each state depending on if they were governed by a Democrat during the encampment (coded as 1) or not (coded as 0).

By taking this coded data from the state voting record and the county voting record, I could create four dummy variables to understand the differing intersections of this independent variable for further analysis. These dummy variables were entered into a subsequent hierarchical regression.

Variable Name	Description	False	True
BlueCounty_BluState	County had more votes for Biden, state had more votes for Biden	0	1
BlueCounty_RedState	County had more votes for Biden, state had more votes for Trump	0	1
RedCounty_BluState	County had more votes for Trump, state had more votes for Biden	0	1
RedCounty_RedState	County had more votes for Trump, state had more votes for Trump	0	1
Blue_Governor	Governor at the time of the encampments was a Democrat	0	1

Table 1: Dummy Variables

4.1.3 Dependent variable

The dependent variable for this cut of data is the existence of an encampment at a university. The existence of a university's encampment was based on local and student journalism at the university. The existence of the encampment was coded as 0 if there was no encampment, and as a 1 if there was an encampment. This data was collected from an exhaustive list of every encampment on wikipedia that was cross referenced by a report from the Harvard Crowd Counting Consortium to ensure that every single encampment at a university was included in the dataset. I then manually coded each university as either having an encampment or note using the Wikipedia and Harvard Consortium list.

4.1.4 Control Variables

In order to eliminate any possible confounding variables, the following variables were selected as control variables: The university's population according to the NAICS, whether the university was private or public, the county poverty rate, the county education rate (defined as the percentage of adults in the county with a bachelor's degree or higher), the county unemployment rate, and the county's median household income according to data from the USDA.

4.1.5 Method of Analysis

A hierarchical regression model was used in IBM's SPSS software. This regression analysis was used to identify if there were significant relationships between a university having an encampment and the coded voting results used as a proxy for local political context. Descriptive statistics on the presence of encampments were also collected.

A regression analysis was used to identify significant relationships between the size of protest and the partisan breakdown of the presidential election results. Additionally, each state's total vote share was collected via the secretary of state of each state. This was used to reduce possible confounders in understanding if there are differences between states in "blue" areas in largely "red" states, vice versa, and more possible ways to further try and isolate confounders to try and find a causal relationship. This also could provide unique marginal cases for further discussion or analysis.

In the hierarchical regression model, the control variables were inserted into the model first, and then in the second model all of the variables including the independent variables of the Biden vote percentage and coded dummy variables were input into the model. This was done to assess the unique contribution the independent variables have and determine whether they are significant.

4.2 Only University Encampments

I then isolated the data to only include the universities that included an encampment. This was done for several reasons. Isolating the data allows for more in depth and clear analysis of the universities that actually had encampments. By isolating the data to only universities that had an encampment I could manually collect additional information regarding the encampments, including whether the encampment ended peacefully or by force. Additionally, this cut of data allows for an analysis of the size of the protests. Finally, this cut of data also allows for comparison between the large N cut of data and this smaller cut to further interrogate the results and determine their strength in either supporting or rejecting the hypothesis.

4.2.1 Independent Variable

For the encampments only cut of data, the independent variable remains the political context, and is operationalized similarly.

4.2.2 Dependent Variable

For the second cut of data, the dependent variable was the number of participants in an encampment. The information regarding the size of the encampment was recorded by manually going through the student newspapers of each university that had an encampment and reading the articles pertaining to the encampment. If no student newspaper was available, local media was substituted. The information on the size of the encampment was also cross-referenced with the

Harvard Consortium dataset. The size of each encampment was coded according to the same methodology employed by McCarthy et al.'s large-N study of student protests during the anti-Vietnam War protests. The following scheme was used to code the data:

Coded Variable	Numerical estimation of participants (if given)	Adjectives used
1	1-9	Small, few, handful
2	10-24	group
3	25-99	Large gathering
4	100-999	Hundreds, mass, mob
5	1,000+0	thousands

Table 2: Coding Scheme for Number of Protestors

In this scheme, preference was given to any numerical estimations, and adjectives and descriptors were only utilized if there was no estimate available.

Outcome of the encampment was coded either as 0, meaning it ended peacefully (including a negotiation, or a decision from the activists to end the encampment), or 1, meaning the encampment was ended forcefully by police.

4.2.3 Control Variables

The control variables for this cut of data are the same from the large cut of data. This means that the poverty rate, education rate, median household income, university population, and whether the school was private or public were all inputted into separate blocks into the hierarchical regression to control. Additionally, counties were coded as "Blue" if Biden received a plurality of the votes, and states were coded as "Blue" if Biden received a plurality of the votes in the state.

4.2.3 Method of Analysis

A hierarchical regression was used to analyze the data. The dependent variable was the coded amount of participation in an encampment according to the scheme initially used by McCarthy et al.

In the first model, the raw vote totals for Trump, Biden, and other candidates in the county were included, as well as coded variables for whether the county had a plurality of votes for Biden and if the state the university resides in had a plurality of votes for Joe Biden. The second block includes the control variables related to the university like if the university was public or private and the university's population. The third and final block of data looks at county statistics of poverty, the unemployment rate, the median household income, and the level of education (defined as the proportion of adults with a bachelor's degree or higher) from the USDA.

4.3 Qualitative Data

Three interviews were conducted over the phone and each lasted approximately 45 minutes. All interviews were completely anonymous and any identifying information was deleted from any transcription. Interviews were recorded using the Otter.ai software, which meets the Soc Type II requirements for confidentiality and privacy. The interviews followed an interview guide (see Appendix A), however follow up questions were occasionally asked to allow interviewees to expand upon their thought process. All interviewees responded to a pre-interview screening and gave oral consent. Interviewees were selected based on an initial professional connection, and then snowball sampling was employed to find other interviewees for this research paper. This process was approved by UCSD's Institutional Review Board.¹

¹ IRB Protocol Number 811917

This study employs a mixed method design, which has been specifically identified as being particularly suitable for understanding protests in the context of political science (Vráblíková 2017, 24). More specifically the qualitative data will serve as a separated complement, where the responses given by the participants of the encampments will help contextualize and evaluate the conclusions drawn from the quantitative data in the conclusion of this paper. The qualitative data will involve snowball sampling for interviews of members of the UCSD encampment via private contacts I have as the News Editor of the student newspaper The UCSD Guardian. The interviews will try to directly understand the decision making process of organizers and rank and file members of the encampments in participating in the demonstration in the context of the local political context of the university. This qualitative data helps establish causality in tandem with the quantitative data, because while the quantitative data may be more objective in its findings, only the qualitative data can have direct engagement with participants of the encampment and have them describe their own motivations and experiences and directly engage with questions pertaining to the hypothesis. These questions include the specific political opportunities and structures and how they relate towards mobilization of the encampments both from the perspective of organizers and participants. Additionally, by interviewing those with similar personal politics but who did not participate in the encampments it could draw clear lines in seeing if political and institutional barriers deterred certain individuals from protesting. While UCSD was chosen in part because of convenience, it could be a great contrasting case to members of a university in a "red" state in a "red" area in trying to understand if that played a role, in comparison to UCSD which is in a "blue" state in a "blue" city and area.

To interpret the data, key themes and common responses will be tabulated. These common responses and themes will then be discussed in depth in relation to what these potential findings mean in relation to the hypothesis.

This qualitative data serves as a separated complement to the quantitative data. This means the results of the qualitative data will be interpreted in conjunction with the results of the quantitative data in the conclusion, to hopefully further corroborate the findings of the quantitative data and truly establish a causal relationship of the overarching hypothesis relating political environments to turnout at the pro-Palestine encampments. Even if the quantitative and qualitative data are not in agreement, this still allows for a rich discussion in trying to understand any such discrepancy which will be illuminated in the conclusion.

5 Data Analysis

5.1.1 National Data Frequencies

First, I collected descriptive statistics for the occurrence of encampments from the dataset. The resulting frequencies are found in the table below.

Description	Frequency
No encampment	1296
Had an encampment	128
Total	1424

Table 3:	Frequency	of University	Encampments
----------	-----------	---------------	--------------------

Of the 1424 universities collected for the dataset, only 128 universities had an encampment. I then further analyzed the frequencies of the differing independent variables I employed on just the universities with an encampment.

Variable	Frequency	Percentage
Blue_County	121	94.5
Blue_State	83	64.8
Blue_Governor	105	82.0
BlueCounty_BlueState	80	62.5
BlueCounty_RedState	41	32.0
RedCounty_BlueState	3	2.3
RedCounty_RedState	4	3.1

Table 4: Encampment only Political Context Frequencies

These frequencies further qualify my hypothesis, considering that 62.5% of encampments occurred in a blue county that's in a blue state, and that 94.5% of encampments occurred in a blue county. Additionally, what is interesting is that despite 64.8% of encampments occurring in states that voted for Joe Biden, 82% of the encampments occurred in states where the governor is a Democrat. This initially does seem to suggest that it is not so much the grassroots support of a candidate, but rather the partisan makeup of institutions that could potentially be impacting whether a protest occurs. In other words, this evidence seems to favor the punishment mechanism over the preference mechanism.

Additionally, I collected the average for the Biden_County_Vote percentage for just the universities with an encampment, and the average was at 63.5%. This is much higher than Biden's national vote percentage of 51.3%. While on its own insufficient to draw any definitive conclusions, these initial frequency results do seem to give further credence to my hypothesis, as it suggests on average the universities that had encampments occurred in counties where Joe Biden had a lot more votes than he did nationally. This leads into the regression, which allows

for a comparison to see if there is a significant relationship between the independent variables of political context and the dependent variable of encampment existence.

5.1.2 National Dataset Hierarchical Regression

The null hypothesis for this cut of data is that there is no significant relationship between the more blue precincts in blue states and the existence of an encampment at a university. The alternative hypothesis is that there is a significant relationship between the blue precincts in blue states and the existence of an encampment at a university. Because dummy variables were used, red precincts in red states were excluded to serve as the reference variable.

The first output from SPSS from the linear regression analysis is the model summary, as seen in the table below.

Model Number	R Squared value
Model 1	0.217
Model 2	0.235

Table 5: Model Summary of all University Dataset

As mentioned in the research design section, the first model includes only the control variables, whereas the second model adds in the independent variables, which includes The Biden County Vote, Blue State, and Blue Governor variables.

The results of the model indicate that there is an increase in the R squared value once the independent variables are introduced. This indicates that the model's accuracy improved with the introduction of the independent variables, which is favorable evidence for my hypothesis.

The next portion of the hierarchical regression is calculating the coefficient and p-values from the hierarchical regression itself.

Model		Unstandardize B	d Coefficients Std. Error	Standardized Coefficients Beta	t	Sig.
1	(Constant)	152	.061		-2.503	.012
	Poverty	002	.003	032	839	.402
	Education	.003	.001	.104	2.889	.004
	Unemployment	.027	.009	.088	3.031	.002
	Median_Household_Incom e	-5.018E-8	.000	003	074	.941
	Private_University	009	.016	015	611	.542
	Uni_Population	8.842E-6	.000	.429	16.508	<.001
2	(Constant)	.016	.069		.233	.816
	Poverty	006	.003	084	-2.045	.041
	Education	.000	.001	009	199	.842
	Unemployment	.004	.010	.013	.398	.691
	Median_Household_Incom e	-1.071E-6	.000	070	-1.504	.133
	Private_University	014	.015	022	883	.377
	Uni_Population	8.957E-6	.000	.434	16.872	<.001
	Biden_County_Vote	.002	.001	.126	3.127	.002
	Blue_Governor	.045	.018	.077	2.469	.014
	Blue_State	.028	.020	.048	1.410	.159

Table 6: Coefficients and P-Values from all University Dataset Regression

The results of both the Blue_Governor and Biden_County_Vote variables are below the critical value of 0.05, meaning that they are significant. Two control variables are also significant, including the poverty variable and the uni population variables.

The coefficient for Biden_County_Vote is positive. This is favorable for my hypothesis given that the results are significant, because it indicates that encampments are significantly more likely to occur the greater the percentage Biden received in the county. This is similar to the Blue_Governor variable, with the coefficient for this variable also being positive, indicating that states governed by Democrats have universities that are more likely to have an encampment.

University population also has a large impact on encampments. Universities with a larger population means there is a higher base of support for the protest to draw on, which goes back to

the idea of resource mobilization impacting protests. This is similar to the poverty variable, which is also significant with a negative coefficient, indicating that the lower the poverty rate the more likely an encampment. The fact that universities in counties with less poverty, and likely more resources, had more encampments further ties back into the general idea of resource mobilization. That said, the results seem much weaker for the control variable of poverty, given that they are not initially significant and only become significant with the addition of the independent variables.

To further interrogate marginal cases and understand the intersection between state politics and local politics, I ran a separate hierarchical regression, simply swapping out the independent variables for the dummy variables outlined above.

Model NumberR Squared ValueModel 10.217Model 20.231

Table 7: Model Summary of all University Dataset with Dummy Variables

The R squared value indicates a similar story to the first version of the hierarchical regression, with the R squared value being slightly smaller with the dummy variables than with the original independent variables. However, this still indicates that the model is improved with these dummy variables.

		Unstandardize		Standardized Coefficients		
Model		В	Std. Error	Beta	t	Sig.
1	(Constant)	152	.061		-2.503	.012
	Uni_Population	8.842E-6	.000	.429	16.508	<.001
	Poverty	002	.003	032	839	.402
	Education	.003	.001	.104	2.889	.004
	Unemployment	.027	.009	.088	3.031	.002
	Median_Household_Incom e	-5.018E-8	.000	003	074	.941
	Private_University	009	.016	015	611	.542
2	(Constant)	.042	.076		.544	.587
	Uni_Population	8.837E-6	.000	.428	16.556	<.001
	Poverty	004	.003	061	-1.605	.109
	Education	.001	.001	.044	1.060	.289
	Unemployment	.012	.009	.039	1.259	.208
	Median_Household_Incom e	-1.068E-6	.000	070	-1.480	.139
	Private_University	012	.015	019	765	.445
	BlueCounty_BlueState	.081	.031	.136	2.652	.008
	BlueCounty_RedState	.009	.031	.014	.291	.771
	RedCounty_RedState	026	.029	043	908	.364

Table 8: Coefficients and P-Values from all University Dataset with Dummy Variables

These results indicate that in the most Democratic context, of BlueCounty_BlueState the results are significant with a positive coefficient, meaning that the encampments are significantly more likely to occur in the political contexts that are the most democratic. The RedCounty_BlueState variable was excluded as there needed to be a reference for the model. Interestingly, the other two dummy variables included in the model are not significant. This may indicate that only when the context is extremely favorable to a movement, both from the punishment and preference mechanism perspectives, does this materialize in some form of impact on the occurrence of protest.

5.2.1 Encampment Only Hierarchical Regression

128 universities had pro-Palestinian encampments. Once I isolated these universities that had an encampment, I could conduct further regressions on the rate of participation in these encampments based off of the coding scheme outlined in the research design. Similar to the universal dataset, the first model in the regression had all of the control variables, while the second model had the Biden_County vote inputted into the model. As the results show in the table below, the model's accuracy increased from an R square value of 0.184 to 0.271 with the addition of the independent variables.

Model Number	R Squared Value
1	0.184
2	0.271

Table 9: Model Summary of Encampment Only Regression Models

		Unstandardize	d Coefficients	Standardized Coefficients		
Model		В	Std. Error	Beta	t	Sig.
1	(Constant)	1.541	.602		2.561	.012
	Uni_Population	8.675E-6	.000	.252	2.725	.007
	Private	.445	.133	.294	3.353	.001
	Median_Household_Incom e	8.128E-7	.000	.021	.150	.881
	Unemployment	.216	.088	.267	2.466	.015
	Education	.015	.009	.187	1.611	.110
	Poverty	016	.027	075	585	.560
2	(Constant)	1.962	.591		3.320	.001
	Uni_Population	9.546E-6	.000	.277	3.099	.002
	Private	.268	.136	.177	1.970	.051
	Median_Household_Incom e	-5.370E-6	.000	136	976	.331
	Unemployment	.104	.090	.128	1.159	.249
	Education	003	.010	041	318	.751
	Poverty	051	.030	243	-1.701	.092
	Biden_County_Vote	.023	.008	.356	2.796	.006
	BlueGovernor	.374	.181	.191	2.073	.040

Table 9: Encampments Only Coefficients and P-Values

This data is highly favorable to my hypothesis as it demonstrates a significant relationship between the percentage of votes received by Joe Biden in a university's county and the number of protestors in an encampment. The only significant variables in the second version of the model are the two independent variables and the university population control variable. Furthermore, the coefficients for both independent variables are positive, which further cements more evidence in support of my hypothesis, as it demonstrates that states with Blue Governors were more likely to have university encampments and universities in counties with a higher percentage of the vote for Biden were more likely to have an encampment.

In sum, the amount of information revealed from the quantitative statistical models likely is enough evidence to reject the null hypothesis across both cuts of data.

5.2.3 Analysis of Force Variable

Although not directly related to the hypothesis, data was also collected on whether encampments ended peacefully or by force. The term force was operationalized to mean any encampment that ended as a result of its physical takedown from those not affiliated with the encampment. The data for force was coded as either a 1 or a 0, with 1 meaning force was used and a 0 meaning no force was used. While the crux of the punishment mechanism relies on perception of risk, it is still useful contextual information to see if these perceptions actually manifested in differing responses dependent on the local political context. Additionally, this analysis opens the door for future research. The resulting frequencies are found below.

 Table 10: Frequencies for Force Variable

Description	Frequency
Force used to end encampment	63
No force used to end encampment	65

The resulting frequency information shows that there was a virtual split between encampments that ended peacefully and that authorities forcefully took down the encampment.

The same control variables were inputted into this hierarchical regression as with the encampment only models. The next step was the model summary. The main difference in these models and the models from the encampments only dataset is the force is substituted in for the encampment size variable. The model summary can be found below.

Model	R Square value
1	0.040
2	0.085

Table 11: Model Summary for Force Variable Regression

Similar to the previous findings, the regression model has a larger R square value with the addition of the independent variables. This indicates that the control variables could be exerting more of an impact on whether force was used than the partisanship.

Standardized Unstandardized Coefficients Coefficients В Std. Error Beta Sig. Model t 1 (Constant) -.135 -.178 .859 .758 Uni_Population 3.183E-6 .000 .842 .135 .404 Private -.042 .169 -.038 -.250 .804 Median_Household_Incom .000 .086 .344 .732 2.301E-6 Unemployment .028 .101 .057 .278 .782 .029 .124 .902 Education .001 .012 Poverty .016 .035 .098 .457 .650 2 1.077 -.934 (Constant) -1.006 .355 Uni_Population 3.240E-6 .000 .137 .777 .442 Private .007 .194 .007 .039 .969 Median_Household_Incom 6.805E-6 .000 .254 .845 .403 Unemployment .007 .123 .013 .053 .958 Education -.002 .017 -.038 -.119 .906 .213 Poverty .034 .042 .818 .418 Biden_County_Vote -.005 .013 -.101 -.389 .699 BlueGovernor .290 .257 .231 1.128 .265 BlueCounty_BlueState .493 .609 .488 .810 .422 BlueCounty_RedState .717 .643 .706 1.115 .271 Blue_State .003 .191 .003 016 .987

Table 12: Coefficients and P-values for Force Variable

Interestingly, no variable has a significant relationship with the usage of force in taking down the encampment. This does seem to suggest that in terms of the operationalized definition

of force, both Democratic and Republican political contexts were just as likely to have an encampment end peacefully or with force.

This opens many avenues for future research. It could be that there is some missing variable that could cause the difference in whether a university responds with force or not. These responses could also be influenced by the individual actors making the decisions on how to respond to the encampments, such as the university administration's makeup.

5.3 Qualitative Data

The results of the quantitative data lead perfectly into the necessity of the qualitative data. Without it, it would be extremely difficult to identify any potential causal relationships, especially given that university population and the blue political context both have statistically significant relationships with the existence and size of the encampments. Thus, the qualitative data can help to parse through whether the local political context is impacting people's decisions to join and protest. Moreover, one of the key goals of the qualitative data is to try and identify which causal mechanisms are likely in play.

After conducting the three interviews, I went through each transcript and coded the information based upon similar and divergent themes brought up in the interviews. The coded information is found below.

Coded Theme	Number of Interviews this theme appeared in
Main fear of participating was fear of repression	3
Would not participate in a hypothetical encampment in a "red" state	2
Felt less fear of repression because they live in a "blue" state	1
Explicitly identified some level of ideological congruence with the Democrats	2
Considered the protest to be very left-wing	2
Joined protest because of personal and political beliefs that align with movement	3
Identified themselves as left-wing on the political spectrum	3
Identified the student body at UCSD as generally center-left	3
Felt like the university was influenced by political pressures in its response to the encampment	3

Table 14: Themes discussed in qualitative interviews

Overall, the results from the three interviews provide further evidence in favor of my hypothesis that the local political context does impact protest ability. Additionally, it gives further insight into which of the two outlined causal mechanisms are having a greater impact. Over the course of the three interviews, the interviewees will be labeled as interviewee 1, 2, and 3 respectively.

All three interviewees described personal moral beliefs and convictions as the motivator for them joining the encampments. Interviewee 1 described these convictions as such: "All my life I wanted to do something to help Palestine." This belief that joining the encampments would help people in Palestine and Gaza was present in all three interviews. While all of the interviewees described differing levels of complaint towards the Democratic party, it was evident that in comparison to the Republican party the Democratic party was at least perceived to be more sympathetic towards the pro-Palestinian movement. However, this higher level of perceived sympathy is nuanced, as interviewee 3 put it: "Democrats are definitely more supportive of the Pro-Palestinian movement. But I also think that Democrats, especially more moderate Democrats haven't been very supportive and kind of have been similar to Republicans in a lot of ways." This articulation is relevant to my hypothesis because the crux of the punishment and preference mechanisms relies upon the idea that the perceptions of risk and support by the encampment members are in some way impacted by the partisan makeup of their university. Even though all interviewees had complaints about the Democratic party, the fact that there was a recognition, particularly by interviewee 3, that the Democratic party is closer to the pro-Palestinian movement's ideals is important because it validates my hypothesis that Democratic political contexts are more likely to have more encampments that are larger in size.

In all the interviews, the interviewee identified the perceived risk of arrest or state force/repression as their primary cost when deciding whether to participate in the movement, and all three said that their main motivation in joining the protest relied on personal and political moral convictions that made them compelled to join the movement. This supports that the rational choice theory is in play, as the participants weighed the risk of being punished against the benefit of participating in a movement that they care about. The fact that two of the interviewees explicitly said that they would not participate in the movement if it took place in a "red" state is perhaps the most direct and clear piece of evidence in support of my hypothesis, because two of the participants in the interviews directly acknowledged that the risks would be
more significant and likely in a "red" state to the extent that they would not join a hypothetical encampment in the first place. This seems to further the punishment mechanism as the primary causal mechanism, as the fear of punishment was articulated explicitly as the main risk by all participants that were interviewed, and two of them specifically linked those perceptions of risk to the state-level partisan makeup. In fact, in interview 2 the participant even explicitly mentioned that they felt that before they joined the movement, because California is a "blue" state, that they felt they had some level of protection, and consciously used that when they were weighing their decisions to join the protest. Moreover, when asked a hypothetical question if they would still participate in an encampment if it occurred in a red state, interviewee 2 responded by saying: "If I were in Tuscaloosa, Alabama I would be way more scared. I would tell my friends and people involved not even to try it. Being in California gives us some form of leeway." This response clearly demonstrates that the regional political context is a major factor in whether an individual will decide to join an encampment to a severe extent. This response strongly bolsters my hypothesis and demonstrates the significant impact the punishment mechanism has in explaining the hypothesis.

Furthermore, as all three interviewees believed that the university faced outward political pressure to use force to end the encampment, it furthers the idea that at the very least, these perceptions of political pressures impact these risk calculations that potential encampment participants must weigh. Given that all interviewees identified themselves as center-left to left-wing, and two explicitly identified some level of ideological congruence with the Democratic party, I argue that this demonstrates a clear connection between the political context and the risk calculation potential encampment participants have to make. The biggest deterrent from their activism being linked to political pressures seems to validate my hypothesis that the

partisan makeup of a given area impacts the decision making process of potential participants of the pro-Palestine movement.

However, there is also evidence in the interviews that can also suggest that the preference mechanism might also be a potential factor, as one participant mentioned that they felt that there would be less support for their movement in a "red" area that would impact the size of a potential encampment. All three also perceived the average UCSD student to be center-left. Moreover, two of the three interviewees claimed that the Pro-Palestine movement at UCSD is explicitly left-wing or leftist. Additionally, all three said that they felt the average student at UCSD was generally supportive of the pro-Palestine movement and their encampment. This does suggest that there was a feeling of general support from the student community.

Despite this evidence in favor of the preference mechanism, I argue that the findings of the interviews do demonstrate that the more prescient mechanism is the punishment mechanism. None of the participants mentioned a feeling of support as a necessary condition for them to join the movement, but all three did mention that the risk to their participation was a fear of some form of repression by either some governmental agency.

6 Conclusion

6.1 Discussion

I argue that the overall results provide substantial evidence in favor of my hypothesis, which is that the local partisan political context impacts the existence and size of the pro-Palestine encampments. More specifically, the data shows that the universities that existed in democratic precincts in democratic states were more likely to have encampments than other universities.

In the universal large N cut of data, the independent variables of the percentage of votes Biden received in the county, if the state was governed by a Democrat, and the dummy variable of "Blue" County and "Blue" state variables all had significant positive relationships with the existence of encampments. That said, while the control variable of university population was also significant, I do not think this alone disproves my hypothesis. This is for several reasons. First, addressing concerns of university population impacting the local political context, if that were the case, then the relationships for the two other variables measuring partisanship would also be expected to be significant across all models, but they are not. This indicates that while the university population may also be causing the existence of these encampments, it is likely an independent causal factor, and not necessarily disproving the potential causality of the local partisanship on the existence of the encampments. Additionally, the poverty variable was also barely significant in the universal dataset. However, this was not a trend continued into the encampments only dataset, which I argue means that poverty is likely not a major causal variable for the outcome of the encampments. I argue that the independent variables of Biden's vote and the Democratic governance of a state are much more likely to be major causal factors because they remain consistent amongst both cuts of data, both the universal and isolated encampment only dataset. The fact that the dummy variable for a Blue county in a Blue state was also significant underscores the fact that the political context, regardless of how it's operationalized as a variable, paints a consistent narrative in favor of the hypothesis across all cuts of data.

The qualitative data provides contextual evidence of a specific case study that further validates my hypothesis. All interviewees acknowledged that punishment was the first and main risk they perceived when deciding whether to participate in the encampment. Additionally, two identified that this risk would become insurmountable and outweigh their calculation of whether

to join if the local partisan context were to become much more Republican. While the qualitative data does give some evidence to both potential causal mechanisms, it is clear that the overlying narrative from the interviewees is one where the predominant causal mechanism is the punishment mechanism.

Another strength of the results is that reverse causality is extremely unlikely, given that the measurements of partisanship were collected before these protests materialized, so the existence of the encampments could not affect the operationalized context of political context. Additionally, while these protests were controversial, there has been no empirical peer-reviewed evidence to suggest that these protests created a massive party shift/ideological shift in the American populace over the course of just a few months. This further builds credibility for the results in favor of the hypothesis.

6.2 Limitations

With any research design centering around a regression analysis, one of the most apparent limitations is the problem of correlation not causation. While this paper includes several control variables, and even qualitative data to try and rectify this problem, this concern is still a potential limitation on the research.

Another potential limitation is the operationalization of the data itself. How people perceive their political context may differ from the actual reality of the political context itself. By measuring voting data as a proxy for the political context, it may miss qualitative judgements by university students on how supportive their political context is towards the pro-Palestine encampments that the voting data does not capture. Moreover, the classification of a political context as a binary "red" or "blue" in certain variables could be seen as failing to capture the variance of political contexts, including the idea of "purple" or swing states, that do not have a dominant partisan ideology and instead tend to swing between the two political parties.

Finally, it is important to mention some of the inherent limitations with qualitative data from the interviews. There is always the potential of research bias, especially in the interpretation of the themes discussed in the interviews. Moreover, the interviews, while in depth and highly informative, were limited in quantity. These interviews were also limited to just one university that exists in one political context, which means the results of the qualitative data on their own are difficult to generalize. Additionally, since the research relied on personal and professional connections, there is further risk that the interviews were not wholly representative. Again, while the interviews were not the crux of the research and many of these concerns are mitigated by the mixed methods design that relies on quantitative data, this is still an important limitation.

6.3 Future Research

What is interesting about these results is the fact that the evidence seems to provide support for political process theory, which recently has fallen out of favor (Meyer 2004). One of the possible reasons that the statistically significant results in favor of my hypothesis despite historically mixed results for political process theory could be because of an increase in political polarization. Partisan political polarization has been increasing in the United States, with the impacts of this increasing polarization impacting even the brands from which consumers purchase on the basis that they view that brand as supportive to their partisan political ideology (Pierson and Schickler 2020; Schoenmueller et. al. 2023). Thus, it stands to reason that as political polarization continues to dominate the individual actions of individuals, this could further galvanize people in certain political strongholds to join protests they find supportive of their political party. It could also be that policymakers, in an extremely polarized environment, feel more concerned with repressing social movements that they view as contrary to their own party, creating a chilling effect on the rates of participation of protests in these local contexts. That said, the results certainly seem to indicate that there may be a need to revisit political process theory in an era of heightened partisan polarization.

Additionally, there are still a plethora of other ways in which to evaluate the data collected from perspectives in political science and even in related fields of political sociology and even social psychology. This could include looking at more recent electoral results, such as the results from the 2024 election. There could also be a more in depth breakdown at demographic information, such as data on race, ethnicity, and even religion could be used to identify if there are other potential causal variables.

Moreover, with the recent actions from the Trump administration in relation to these protests, it further raises questions regarding the differential treatment exerted on protests and protestors by differing political parties (Allen 2025). It will be important in the future to evaluate whether these increasingly sharp attacks on political enemies' protests manifest in an impact on the existence of the protests and their own size. Another portion of additional research would also be to further scrutinize and evaluate the subsequent action taken by governmental forces after a demonstration occurs. While this paper briefly mentions the difference in protests ending in peace or by force, future research could center this question as the main hypothesis. This is an especially prominent possible focus for future research given that the results found in this paper fail to find any strong evidence of a relationship existing between any of the collected variables and the usage of force/peace in ending the encampments.

Finally, there is a large amount of future research that could be done through further qualitative analysis of the members of these encampments. Interviewing participants of

encampments at other universities, particularly those in differing political contexts in "red" states, would be highly insightful to further evaluate my hypothesis.

While this is an extremely contentious political problem, political science should not shy away from covering issues because they are controversial or personal. If anything, it is imperative to try and objectively assess phenomena within the field of political science to make sense of new and often unique situations that arise in contemporary politics. It is with this goal in mind that this paper seeks to contribute knowledge to the field of political science.

Appendices

Appendix A: Interview Guide

The interviewee

- 1. Did you participate in the Pro-Palestine encampment?
 - a. In what capacity? What was your role?
- 2. What was your general experience participating in the encampment?
- 3. What motivated you to participate in it?
- 4. When did you decide to join the encampment? What explained this timing?
- 5. Did you ever consider leaving the encampment?
- 6. What were the main risks you considered when deciding whether to join the encampment?
 - a. How did you assess them?
- 7. What do you think the goal of the encampment was?
 - a. Did you agree with it fully?
- 8. When you got involved, what did you expect the response of the admin to be? a. Why?
- 9. On a 10 point ideology scale, with 0 being far left and 10 being far right, where would you place yourself?
- 10. Did the fact that UCSD is in a blue state shape your perceptions of risk at all?
- 11. Did the fact that UCSD is in a large, mostly progressive city shape your decision to participate?
- 12. Have you ever been involved in political activities before this?
 - a. What kind?
- 13. Have you voted in national or local elections?
 - a. Which ones?
- 14. Did this experience change you in terms of your politics?
- 15. Hypothetical: if you were a student at a similar university but one in a "red" state, do you think you would have joined? why/why not? Do you think the protest would have occurred at all? why/why not?
 - a. Would go way up not even try it. Some sort of leeway
 - b. What if you were in a "red" city or local area?

Other participants

- 1. What do you think motivated other participants to participate?
- 2. How would you describe fellow encampment members politically?
- 3. On a 10 point ideology scale, with 0 being far left and 10 being far right, where would you place other encampment members?
- 4. Do you think the leadership was distinct in terms of their politics from rank and file members of the encampment?
 - a. How so?
- 5. Why do you think leadership decided to have a protest at UCSD?
 - a. What do you think they expected the result to be?
- 6. Were participants primarily from campus?

a. Were there any participants not on campus?

Campus

- 1. Do you think other students on campus that did not join the encampments supported your protest?
 - a. Did that support impact your ability to have an encampment and sustain it?
- 2. On a 10 point ideology scale, with 0 being far left and 10 being far right, where would you place the average student at UCSD?
- 3. Did you feel that your protest was supported by different student groups on campus?
 - a. Was their support important to your protest?
 - b. Did you face any opposition from any organizations or students on campus?
 - i. On a 10 point ideology scale, with 0 being far left and 10 being far right, where would you place the average student/organization that opposed your protest? Why?
 - c. Do you think a lot of groups and students were neutral or indifferent to the protest? Why?
- 4. Did you become aware of more student groups when participating?
- 5. How would you describe the participating student groups politically? Were they from the same side of the political spectrum, or were there a mix across the political spectrum?

Campus admin

- 1. How did your campus administration respond to the encampment?
 - a. If the campus was in a red state [red city] do you think the campus administration would have reacted the same way? Why?
- 2. Why do you think campus administration acted as they did?
- 3. Do you think the campus administration was politically motivated?
 - a. How so?
- 4. On a 10 point ideology scale, with 0 being far left and 10 being far right, where would you place the members of campus administration at UCSD?
- Do you think campus administration faced pressures to react the way they did?
 a. From whom?
- 6. Do you think that the administration had the best interests of the students in mind during their response to the encampments?
 - a. (If not), whose interests did they have in mind in your opinion?

Thank you! This study was about the politics associated with different campus political environments in relation to how universities responded to Pro-Palestine encampments. Knowing this now, is there anything you wish to add? How did you feel?

Thank you for your responses.

Appendix B: Quantitative Dataset

						lueCounty BlueStateBl	ueCounty RedStateRed					cationUn			IncomePrivate Universi
	0	0	35.89	0	0	0	0	0	0	0	5433	18.2	31.9	3.6	46,433
0	0	0	26.46	0	0	0	0	0	0	0	6335	13.3	29	3.3	61,798
	0	0	85.26 48.14	0	1	1	1	0	0	0	10505	11.9 21.7	60.1 29.3	3.3 3.7	135,366 53,942
	5	0	48.14 54.11	0	1	1	1	0	0	0		5.8	48.7	3.7	135,528
	5	0	74.22	0	1	1	1	0	0	0	9312 1944	13.2	41.9	4.4	76,614
)	0	39.13	0	0	1	0	0	0	0		10.6	22.3	4.4	67,486
	5	0	60.85	0	0	0	0	0	0	0	2285	12.4	38.4	4.2	72,129
		0	60.85 83.09	0	0	0	0	0	0	0	2259 1468	12.4	38.4 46.7	2.8	72,129
	5	0	53.31	0	0	0	0	0	0	0	1074	14.1	33.2	1.8	66,952
	5	0	39.26	0	0	0	0	0	0	0	1129	9.9	41.3	3.2	82,769
	5	0	64.55	0	1	1	1	0	0	0	1450	12.9	45.2	3.3	76,997
	5	0	64.55	0	1	1	1	0	0	0	3103	12.9	45.2	3.3	76,997
		0	69.62	0	0	0	0	0	0	0		26.4	23.2	4.3	42,629
	5	0	83.29	0	1	1	1	0	0	0		27.7	22	6.8	45,864
	5	0	58.05	0	1	1	1	0	0	0	1730	12.5	41.1	3.9	84,615
	5	0	43.57	0	0	1	0	0	0	0	2005	14	22.2	4.3	59,652
Ċ		0	45.08	0	0	1	0	0	0	0	2005	11.9	27.2	3.5	72,157
	5	0	29.1	0	1	1	1	0	0	0	2016	15	24.6	4.3	55,466
	5	0	30.69	0	0	1	0	0	0	0	2143	12.6	21.2	3.7	54,774
		0	60.21	0	1	1	1	0	0	0	4023	10.1	42.1	3.9	98,365
	0	0	34.95	0	0	1	0	0	0	0	1781	16.3	19	4.4	58,074
	5	0	45.08	0	0	1	0	0	0	0	2984	11.9	27.2	3.5	72.157
	5	0	69.07	0	0	1	0	0	0	0	2319	17.1	33.9	3.8	58,375
-		0	64.89	0	0	0	0	0	0	0		13.8	34.8	3.8	70.871
-		0	63.35	0	0	0	0	0	0	0	8517	15.6	34.1	3.3	62,776
	5	0	63.35	0	0	0	0	0	0	0	8517	15.6	34.1	3.3	62,776
	5	0	62.41	0	0	1	0	0	0	0	5969	6.9	52.3	2.8	106,743
	5	0	50.13	0	0	1	0	0	0		17069	11.1	35.9	3.4	83,668
_		0	72.57	0	0	0	0	0	0		1570	13	58	3.4	89,798
	5	0	57.73	0	1	1	1	0	0	0	3419	17.6	28.9	4.3	64,030
_		0	86.42	0	1	1	1	0	0	0	2172	16.5	39.6	4.6	95,514
	5	0	43.77	0	0	0	0	0	0		52455	14.8	33.7	2.7	87,259
	5	0	79.55	0	1	1	1	0	0	0	2973	10.5	55.6	3.3	87,619
	5	0	92.15	0	1	1	1	0	0		17561	15.2	63.6	4.9	99,897
	, 	0	72.12	0	1	1	1	0	0	0	2767	12.4	51.1	2.9	81,878
	5	0	28.27	0	0	0	0	0	0	0	4760	13.5	26.4	2.8	62,708
	5	0	60.16	0	0	0	0	0	0	0	1782	13	20.1	3.6	58,926
	5	0	60.16	0	0	0	ő	0	0	0	1782	13	20.1	3.6	58,926
-		0	45.34	0	0	1	0	0	ő	0	3967	13.6	30.2	4.2	61,790
-		0	27.07	0	0	0	0	0	0		11732	14.6	25.1	3.5	60,348
	5	0	57.58	0	1	1	1	0	0	0	1806	10.9	39	3.5	86,078
	5	0	71.03	0	1	1	1	0	0	0	1118	13.7	35.5	5	82,455
-	5	0	57.52	0	1	1	1	0	0		1067	9	35.3	2.3	76,227
	0	0	53.14	0	0	1	0	0	0		23259	18.6	47.7	3	57,888
	5	0	51.91	0	ő	1	0	0	0	0	1913	10.5	39.6	3.2	76,923
-		0	62.41	0	0	1	0	0	0	ő	4257	6.9	52.3	2.8	106,743
	5	0	50.13	0	ő	1	0	0	0	ő	1188	11.1	35.9	3.4	83,668
	5	0	50.13	0	0	1	0	0	0		1044	11.1	35.9	3.4	83,668
	1	4	50.13	0	0	1	0	0	0			11.1	35.9	3.4	83,668
-	1	0	50.13	0	0	1	0	0	0			11.1	35.9	3.4	83,668
	5	0	50.13	0	0	1	0	0	0		10608	11.1	35.9	3.4	83,668
	5	0	50.13	0	0	1	0	0	0	0	9752	11.1	35.9	3.4	83,668
		0	30.95	0	0	0	0	0	ő	0	3663	16.1	27.8	2.7	60,580
	0	0	23.62	0	0	0	0	0	0	0	15053	13.5	24.6	3.9	57,881
	5	0	71.03	0	1	1	1	0	0	0	11954	13.7	35.5	5	82,455
	5	0	32.6	0	0	1	0	0	0	0	2981	10.3	32.4	3.5	69,216
	5	0	32.6	0	0	1	0	0	0	0	1991	10.3	32.4	3.5	69,216
	5	0	60.21	0	1	1	1	0	ō	0	2175	10.1	42.1	3.9	98,365
	5	0	24.77	0	0	0	0	0	0	0	33440	13.5	22.1	3.4	62,037
	5	0	57.58	0	1	1	1	0	0	0	5568	10.9	39	3.5	86,078
	5	0	70.46	0	1	1	1	0	0	0	2978	10.1	53.3	2.6	89,418
	5	0	67.89	0	0	0	0	0	0	0	4025	22.2	24.1	4.6	50,079
	5	0	54.81	0	1	1	1	0	0	0	14608	14.8	24.1	5.1	62,652
	5	0	43.85	0	0	0	0	0	0	0	3047	9.7	34.6	1.8	75,623
	5	0	56.14	0	1	1	1	0	0	0	2476	8.4	37.5	5.1	93,475
	5	0	24.39	0	0	0	0	0	0	0	7131	10.9	23.7	3.7	66,141
	5	0	42.3	0	0	0	0	0	0	0	1625	12.6	31.7	3.7	67,478
	5	o	37.27	0	0	0	0	0	ő	0	11626	10.5	39.5	2.9	81,821
	5	0	59.82	0	0	0	0	0	ō	0	1334	12	33.5	3.4	62,509
	0	0	59.82	0	0	0	0	0	0	0	1761	12	33.5	3.4	62,509
	0	0	71.03	0	1	1		0	0	0	1761	13.7	35.5	5	82,455
(0	0	67.03	0	1	1	1	0	0	0	10969	7.7	57.6	3.1	115,770 67,482
	0	0	68.04	0	0	1	0	0	0	0	4222	13.3	52	2.6	67,482 60,808
	0	0	66.36	0	0	0	0	0	0	0	2760	16	35.9	3.8	60,808 54,059
	0	0	42.35	0	0	0		0	0	0	4293	19.4	25.7	3.8 3.8	54,059
	0	0	42.35	0	0	0	0	0	0	0	25208	19.4	25.7	4.3	61,452
	0	0	64.42	0	0	0	0	0	0	0	25208	18.2	34.2		61,452 90,740
	0	0	53.89	0	1	1	1	0	0	0	1032 3385	8	39.7	3.3 4.6	95,514
(1	5	86.42	0	1	1	1	0	0	0		16.5	39.6	1.8	66,952
(1 0	0	53.31	0	0	0	0	0	0	0	3696 8728	14.1	33.2	4.9	51,535
0			50.95	0	0	1	0	0	0	0	8728	18.8	20.3	3.4	116,044
0	5	0	74.95	0	1	1	1	0	0	0	1445	8.8	55.9	3.4	62,848
))	0		0	1	1	1	0	0	0	2752	13	23.4	4.3	64,030
	0 0 0	0	47.04		1	1	1	0	0	0	3937	17.6	28.9	4.3	68,748
	0 0 0	0 0 0	57.73	0		0	0	0	0	0	10496	16	33.7	3.7	63,160
		0 0 0 0 0	57.73 55.94	0	0			0	0	0	10496 22706	16.7	27.8	3.1	47,479
		0 0 0 0	57.73 55.94 37.49	0	0	0	0					04			
			57.73 55.94 37.49 73.4	0 0 0	0	0	0	0	0	0		21	31.1	3.9	64,719
		0 0 0 0	57.73 55.94 37.49	0	0	0			0	0	5756	16.2	36.6	3.9 3.9	64,719 64,719
		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	57.73 55.94 37.49 73.4 58.87 58.87	0 0 0 0	0 0 0 0 0	0 0 1 1	0 0 0	0 0 0	0	0	5756 4087	16.2 16.2	36.6 36.6		64,719 98,619
			57.73 55.94 37.49 73.4 58.87	0 0 0 0 0	0 0	0 0 1 1 0	0	0	0	0	5756 4087 4087	16.2 16.2 4.9	36.6 36.6 41.9	3.9	64,719 98,619 65,430
		0 0 0 0 0 0 0 0 0 0 0 0	57.73 55.94 37.49 73.4 58.87 58.87 42.84 35.46	0 0 0 0 0 0	0 0 0 0 0	0 0 1 1 0 1	0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0	0 0 0	5756 4087 4087 14569	16.2 16.2 4.9 12.5	36.6 36.6 41.9 26	3.9 2.2 3.4 3.4	64,719 98,619 65,430 67,430
) () () () () () () () () () () () () ()		0 0 0 0 0 0 0 0 0	57.73 55.94 37.49 73.4 58.87 58.87 58.87 42.84 35.46 54.66	0 0 0 0 0	0 0 0 0	0 0 1 1 0 1	0 0 0	0 0 0	0 0 0	0 0 0 0	5756 4087 4087 14569 1771	16.2 16.2 4.9 12.5 11.2	36.6 36.6 41.9 26 25.7	3.9 2.2 3.4	64,719 98,619 65,430 67,430 59,882
			57.73 55.94 37.49 73.4 58.87 58.87 42.84 35.46 54.66 47.24	0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 1	0 0 1 1 0 1 1 1	0 0 0 0 0 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0		5756 4087 4087 14569 1771 1307	16.2 16.2 4.9 12.5 11.2 17.1	36.6 36.6 41.9 26 25.7 32.3	3.9 2.2 3.4 3.4	64,719 98,619 65,430 67,430 59,882 61,010
		0 0 0 0 0 0 0 0 0 0 0 0	57.73 55.94 37.49 73.4 58.87 58.87 42.84 35.46 54.66 47.24 68.4	0 0 0 0 0 0 0 0	0 0 0 0 0 0 0	0 0 1 1 0 1 1 1 0	0 0 0 0 0 0 1	0 0 0 0 0 0 0 0 0 0 0	0 0 0 0		5756 4087 4087 14569 1771 1307 5137	16.2 16.2 4.9 12.5 11.2 17.1 15.9	36.6 36.6 41.9 26 25.7 32.3 41.1	3.9 2.2 3.4 3.4 3.5	64,719 98,619 65,430 67,430 59,882 61,010 59,085
			57.73 55.94 37.49 73.4 58.87 58.87 42.84 35.46 54.66 47.24	0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 1	0 0 1 1 0 1 1 1	0 0 0 0 0 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0		5756 4087 4087 14569 1771 1307 5137 2055	16.2 16.2 4.9 12.5 11.2 17.1	36.6 36.6 41.9 26 25.7 32.3	3.9 2.2 3.4 3.4 3.5 3	64,719 98,619 65,430 67,430 59,882 61,010
			57,73 55,94 37,49 73,4 58,87 58,87 42,84 35,46 54,66 47,24 68,4 31,71 57,66	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 1	0 0 1 1 1 1 1 1 1 0 1 1	0 0 0 0 0 0 1	0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0		5756 4087 4087 14569 1771 1307 5137 2055 2628	16.2 16.2 4.9 12.5 11.2 17.1 15.9 13.6 6.6	36.6 36.6 41.9 26 25.7 32.3 41.1 22.9 51.4	3.9 2.2 3.4 3.4 3.5 3 3.4	64,719 98,639 65,430 67,430 59,882 61,030 59,085 102,383 69,249
) () () () () () () () () () () () () ()			57.73 55.94 37.49 73.4 58.87 58.87 42.84 35.46 54.66 47.24 68.4 31.71	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 1 0 0 0	0 0 1 1 0 1 1 1 0 1	0 0 0 0 0 1 0 0	0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0		5756 4087 4087 14569 1771 1307 5137 2055	16.2 16.2 4.9 12.5 11.2 17.1 15.9 13.6	36.6 36.6 41.9 26 25.7 32.3 41.1 22.9	3.9 2.2 3.4 3.4 3.5 3 3.4 3.4 3.4	64,719 98,619 65,430 67,430 59,882 61,010 59,085 102,383

	Protestore	en County VoteBlue Cou	tyBlue Stat-D	lue GovernorBlu	County BlueState	County Red State Do 10	ounty BlueStateRedCo	Inty RedStatella	nulation	RateEducation	Inemplor	nentMedian House	ehold IncomePrivate Universi
0	# ProtestorsBide	43.06	0		eCounty BlueStateBlue	0 0			opulationPoverty 0 903		Unemployn 32.3	nentMedian Hous 2.9	71,237
0	0	27.75	0	0	0	0 0			0 362	3 12	28	3.1	53,862
0	0	27.75	0	0	0	0 0	0 0		0 140	7 12	28	3.1	53,862
0	0	47.74	0	0		0 0			0 1981	1 21	30.8	4.5	54,642
0	0	43.78	0	0		0 0			0 113		39.6	2.7	81,295
0	0	39.26	0	0		0 0			0 449	9.9	41.3	3.2	82,769
0	0	43.32	0	1		1 0			0 1271		35.3 27.4	5.1	71,737
0	0	36.59 26.07	0	0		0 0			0 175 0 266		35.2	4.5	57,975 52,621
0	0	51.11	0			0 0			0 266 0 108		33.1	3.1	70,203
0	0	72.57	0			0 0			0 133		58	3.4	89,798
0	0	57.66	0	1		1 0					51.4	3.4	102,383
0	0	60.22	0	1		1 0			0 102		45.1	4.1	102,073
0	0	55.94	0	0		0 0			0 155		33.7	4.3	68,748
0	0	62.51	0			1 0			0 259		37.7	2.7	96,304
0	0	75.78	0	1	0	1 0	0		0 480		55.8	1.6	86,579
1 0		53.48	0	1	1	1 0	0		0 1195		43.4	3.6	106,047
0	0	71.03	0	1	1	1 0	0 0		0 149		35.5	5	82,455
0	0	55.51	0	0	0	0 0	0 0		0 393		49	2.4	79,969
0	0	50.95	0	1	1	1 0	0 0		0 246		33	4.3	91,149
0	0	63.06	0	1		1 0			0 196		41.7	3.8	84,551
0	0	59.43	0	0		0 0					44.8	3.3	71,973
0	0	81.21	0	0	1	0 0			0 183		34.6	4.2	56,385
0	0	74.22	0	1		1 0			0 330		41.9	4.4	76,614
0	0	66.74	0	0		0 0			0 128		16.7	4.7	43,871
0	0	64.42	0			0 0			0 234		34.2	4.3	61,452
0	0	55.51	0			0 0			0 448		49	2.4	79,969
0	0	74.95	0	1		1 0			0 254		55.9 19.4	3.4 4.4	116,044 42,209
0	0	66.24 71.03	0	0	-	0 0			0 248		19.4 35.5	4.4	42,209 82,455
0			0	1		1 0				10.7	35.5	5	82,455
0	0	71.03 23.96	0	0		0 0	0		0 177 0 513		23.3	4.2	57,844
0	0	72.57	0			0 0			0 513 0 446		23.3	4.2	89,798
0	0	57.58	0	1		1 0			0 446 0 430		39	3.4	86,078
0	0	47.58	0			0 0			0 430 0 117		32.9	2.9	73,492
0	0	54.37	0			0 0) n		0 133	010	42.2	2.6	74,424
0	0	43.11	0	1		1 0	0		0 489		24.6	4.4	59,451
ő	0	84.94	ő			0 0	o o		0 786		21.3	4.1	57,625
0	0	23.71	0			0 0	0		0 3199		27.3	2.9	55,333
0	0	66.36	0	0	0	0 0	0		0 1760	16	35.9	3.8	60,808
0	0	32.92	0	0	0	0 0	0 0		0 1162		26.6	3.4	61,375
0	0	29.38	0	0	0	0 0	0		0 390	5 19.9	15	3.3	48,245
0	0	55.61	0		0	0 0	0 0		0 176	5 9.9	34.8	3.5	74,134
0	0	46.95	0	0	0	0 0	0 0		0 132	2 22.3	19.7	3.3	50,427
0	0	48.57	0	1	1	1 0	0 0		0 297	5 11.5	30.9	2.6	63,191
0	0	53.85	0	1	0	1 0	0 0		0 121	5 7.7	38.1	1.9	84,898
0	0	43.5	0	1	1	1 0	0 0		0 412		28	3.7	69,851
0	0	55.51	0	0	0	0 0	0 0		0 1224		49	2.4	79,969 45.864
0	0	83.29	0	1	1	1 0	0		0 322		22	6.8 4.3	45,864 64,030
0	0	57.73	0	1	1	1 0) 0		0 166 0 211		28.9 28.6	4.3	84,738
0	0	37.58 54.37	0	1	0	0 0	0		0 211 0 131		42.2	2.6	74,424
0	0	41.98	0	1	1	1 0	, 0		0 1459	. 11.5	35.8	4.9	91,582
0	0	57.58	0	1	1	1 0			0 405	10.1	39	3.5	86,078
0	0	20.27	0		0	0 0			0 405 0 188	10.3	25.4	4.9	53,009
0	0	57.88	0	1	1	1 0	,		0 890	13.0	50.7	2.9	102,413
1	4	42.75	0	1	1	1 0	, o		0 296	1.4	41.1	3.3	82,248
0 0	0	34.8	0	1	1	1 0	, ,		0 1013	/.3	30.6	3.5	69,578
0	0	57.88	0	1	1	1 0			0 816	1 11.0	50.7	2.9	102,413
0	0	49.57	0	1	1	1 0	0		0 671	1.4	25	4.4	58,314
0	0	56.22	0	1	1	1 0	0		0 4014	10.0	51.7	2.8	88,338
0	0	42.75	0	1	1	1 0	0		0 3026	10.2	41.1	3.3	82,248
0	0	39.39	0	1	1	1 0	0		0 233	7.0	23.9	5	70,108
0	0	68.4	0	0	0	0 0	0		0 148		41.1	3	61,010
0	0	54.82	0	0	0	0 0	0		938		51	2.5	63,981
0	0	74.22	0			1 0	0		0 816	5 13.2	41.9	4.4	76,614
0	0	68.4	0	0	0	0 0	0		0 250	9 15.9	41.1	3	61,010
1	5	86.42	0	1		1 0	0 0		0 5099	5 16.5	39.6	4.6	95,514 69,762
0	0	64.68	0			0 0			0 128	4 15.1	42	4.2	53,740
0	0	61.4	0			0 0			948	20.7	29.7	4.2	43,917
0	0	22.1	0			0 0			0 208		20.7	3.4	108,037
0	0	67.57 50.74	0			1 0			0 146	9.3	52.5 39	2.1	79,357
0	0		0			1 0			0 264	13.1	39 55.5	3.3	95,151
0	0	71.41 72.44	0	0		0 0			0 264 0 146		55.5	3.3	80,941
0	0	72.44 74.22	0	1	1	1 0			0 146 0 717	14.6	41.9	4.4	76,614
0	0	53.48	0		1	1 0			0 717	10.2	41.9	3.6	106,047
0	0	26.5	0			0 0			0 463 0 362	0.2	43.4	2	77,778
0	0	79.21	0	1		1 0			n	0.0	48.6	3.6	79,432
0	0	71.5	0			1 0			0 545		45.5	2.7	75,041
0	0	44.13	0	0		0 0			0 616		50.4	2.6	97,778 80,159
0	0	56.92	0	1		1 0			0 654		34.6	3.6	80,159 58,535
0	0	35.6	0	0	0	0 0	0 0		0 249		25.7	2.9	95,514
0	0	86.42	0	1	1	1 0	0 0		0 160		39.6	4.6	95,514 54,735
0	0	87.28	0	1	1	1 0	0 0		0 132		35.4	2.9	54,735 70,861
0	0	48.86	0			1 0			290	·	26	3.7	70,861 74,134
	0	55.61	0			0 0			0 141	•	34.8	3.5	74,134 72,025
0	4	73.51	0			1 0			0 128 3468		59.2	3.1 2.6	59,708
1	0	16.93	0			0 0				•	19	2.6	55,751
1 0 0		38.71	0		0	0 0			0 113		32.4	2.6	74,424
1 0 0	0	54.37	0			0 0			110		42.2	2.6	117,602
1 0 0 0 0	0		0			1 0			0 1100		50.5	3.3	90,740
1 0 0 0 0 0	0	46.37			1	1 0	0		0 169		39.7		57,053
1 0 0 0 0 0 0	0	53.89	0							. 8		2.9	
1 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0	53.89 21.3	0	0	0	0 0			0 383		16.5		88,462
1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	53.89 21.3 29.94	0	0	0	0 0	0 0		0 383 0 122	15.3	16.5 37.2	2.9 2.7 4.6	88,462 95,514
1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	53.89 21.3 29.94 86.42	0 0 0	0 0 1	0 0 1	0 0 0	0 0		0 383 0 122 0 309	15.3 6.6 2 16.5	16.5 37.2 39.6	2.7 4.6 5.5	88,462 95,514 73,244
1 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0	53.89 21.3 29.94 86.42 76.78	0 0 0	0 0 1 1	0 0 1	0 0 0 0 1 0 1 0	0 0 0 0 0 0		0 383 0 122 0 309 0 2212	15.3 6.6 16.5 16.5 7 19.1	16.5 37.2 39.6 41.3	2.7 4.6 5.5 4.6	88,462 95,514 73,244 95,514
1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	53.89 21.3 29.94 86.42	0 0 0	0 0 1 1 1 1	0 0 1 1 1	0 0 0) 0 0 0 0 0 0		0 383 0 122 0 309 0 2212	15.3 6.6 16.5 7 19.1	16.5 37.2 39.6	2.7 4.6 5.5	88,462 95,514 73,244

	d# ProtestorsBiden Co		tyBlue StateBlu	e GovernorBlue	County BlueStateBlue					verty RateEd		nemployme	entMedian Househol	d IncomePrivate Universi
0	0	86.42	0	1	1		0	0		27203	16.5	39.6	4.6	95,514
0	0	86.42	0	1	1		0	0		17769	16.5	39.6	4.6	95,514
0	0	83.29	0	1	1			0		16943	27.7	22	6.8	45,864
0	0	76.78	0	1	1			0	0	6059	19.1	41.3	5.5	73,244
0	0	76.78	0	1	1		0	0		17642	19.1	41.3	5.5	73,244
0	0	72.03	0	1	1			0		22053	13.8	35.3	4.6	80,180
0	0	72.03	0	1	1		0	0	0	8801	13.8	35.3	4.6	80,180
0	0	67.03	0	1	1	1	0	0	0	2989	7.7	57.6	3.1	115,770
0	0	49.66	0	1	1	1	0	0	0	3231	14.1	36.9	3.8	69,888
0	0	56.46	0	1	1	1	0	0	0	3075	14.1	36.9	3.8	69,888
0	0	35.21	0	0	0	D	0	0	0	3551	9.4	34.7	1.9	74,668
0	0	31.32	0	0	0	0	0	0	0	1137	9.4	25	1.7	63,753
0	0	64.89	0	0	0		0	0	0	2739	13.8	34.8	3.8	70,871
0	0	29.05	0	0	0	0	0	0	0	5197	15.5	19.3	3.9	61,680
1	1 4	61.29	0					0		10253	9.4	44.8	1.9	81,205
0	0	51.91					0	0	0	7148	10.5	39.6	3.2	76,923
0	0	66.68		0			0	0	0	2910	10.2	48.6	3.4	80,645
0	0	51.19	0	1			0	0	0	5657	11.7	26.7	4.5	71,866
0	0	51.53		0			0	0	0	2775	6.7	43.9	4.5	
0	0	64.78					0	0	0		38.7	26.6		105,202
0	0	35.02					0	0	0	3631 3082	9.7	29.3	3.6 3.1	39,984
0	0	79.55	0				0	0	0		10.5	55.6		76,596
				1				0		1230			3.3	87,619
1	1 4	74.22	0	1	1		0	0		25295	13.2	41.9	4.4	76,614
0	0	23.74						0	0	2399	10.8	19.9	3.2	68,301
0	0	23.74					0	0	0	2399	10.8	19.9	3.2	68,301
0	0	56.52		-			0	0	0	1901	11.5	39.5	2.9	77,369
0	0	53.05	0	0	1		0	0	0	3985	12.1	33.4	3.6	72,590
0	0	86.42	0	1			0	0	0	1318	16.5	39.6	4.6	95,514
0	0	54.2	0	1			0	0	0	2239	13.2	22.9	4.7	78,779
0	0	83.09	0	0	0	D	0	0	0	1138	13.3	46.7	3.3	76,736
0	0	74.22	0	1	1		0	0		22110	13.2	41.9	4.4	76.614
0	0	43.78	0	0	1	0	0	0	0	2772	8.2	39.6	2.7	81,295
0	0	16.6					0	0	0	1699	9.6	27.2	1.5	81,295 86,963
0	0	74.95	0	1	1		0	0	0	1370	8.8	55.9	3.4	86,963
0	0	66.05		0	1		0	0	0	1679	29.6	32.6		
0	0	22.52					0	0		13327	10.6	33.3	7.9	40,632
0	0	34.36			0		0	ō	0	2810	10.6	20.1	2.8	75,572
0	0	34.36 50.3	0	1	1		0	0	0		10.5	42.1	2.4	72,948
			0		1			0		2126			3.1	99,839
0	0	74.22					0	0	0	3900	13.2	41.9	4.4	76,614
0	0	82.33	0				0	0	0	2369	8.7	60.5	3.3	135,960
0	0	15.85			0		0	0	0	1979	6.9	30.4	1.9	75,994
0	0	56.52		0	0		0	0	0	5952	11.5	39.5	2.9	77,369
0	0	51.14	0	1	1		0	0	0	2707	4.8	57.2	3.7	131,562
1	0 4	81.21				D	0	0		28508	20.3	34.6	4.2	56,385
0	- 0	38.71	0	0	0	0	0	0	0	2032	14.1	32.4	2.6	55,751
1	1 4	80.42	0	0	1	0	0	0	0	35280	10.6	53.5	3.1	79,524
0	0	59.43	0	0	1	0	0	0	0	10939	11.6	44.8	3.3	71,973
0	0	34.41	0	0	0	0	0	0	0	1113	17.3	20.2	3.4	51.622
0	0	34.41	0	0	0	0	0	0	0	1113	17.3	20.2	3.4	51,622
0	0	53.96	0	0	1	0	0	0	0	34648	19.4	33.9	3.9	57,319
0	0	26.87					0	0	0	4109	13.9	29.7	3.1	56,077
0	0	52.43			1		0	0	0	6488	9.6	27.9	4.5	79,824
0	0	30.96			0		0	0		16253	16.3	33.5	3.3	64.270
ő	0	26.61					0	0	0	2031	17.2	22.5	4.7	64,636
ő		46.79			4		0	0	0		12.4	34.6	3.9	80,159
	0		0					0		5587				
0	0	35.59	0	1	1		0	0	0	9861	17.3	27.1	4.4 3.9	57,712
0	0	35.45		0			0	0		16828	13.8	34.2		60,994
0	0	72.44		0	1		0	0		18301	14.6	58.1	3.3	80,941
0	0	27.27	0	1	1		0	0	0	5825	19.3	21.3	3.6	47,854
0	0	28.47	0	1	1		0	0	0	3285	13.6	24.9	3.9	57,091
0	0	62.75		0	1		0	0	0	4167	10.8	41.6	3.2	83,856
0	0	45.96	0		1		0	0		13716	12.5	32.1	4.2	69,079
0	0	49.44		0	0		0	0	0	2228	11.3	35.8	2.8	66,427
0	0	75.46	0	0	1	D	0	0	0	2455	10	54.4	2.3	85,189
0	0	49.66		0	1		0	0	0	4904	14.5	29.6	3.8	56,489
0	0	51.11		0	0		0	0	0	2511	14.5	33.1	3.1	70,203
0	0	49.41					0	0	0	2390	15.5	26.9	4.1	59,066
0	0	41.17		0			0	0	0	2419	8.6	31	2.7	82,434
0	0	57.66		1	1		0	0	0	4120	6.6	51.4	3.4	102,383
0	0	42.21	0	1	1		0	0	0	1032	16.9	26.9	4	56,664
ō	0	45.1		0	1		0	0	0	8707	12.5	27.7	3.4	61.034
0	0	34.53					0	0	0	3522	12.8	35.9	3.6	64,187
0	0	42.38						0		3522 14797	11.9	27.2	3.3	64,601
1	4	42.36 80.64	0		1		0	0	0	6324	15.1	49.9	6.2	84,738
0	1 0	80.64						0	0	6324 2351	15.1	49.9	3.2	84,738 84,548
0											15.1	49.9	3.2	
	0	14.8						0	0	1087				49,940
1	4	83.09					0	0		26950	13.3	46.7	3.3	76,736
0	0	43.1		-			0	0	0	6575	13.7	27.7	2.6	55,863
0	0	63.44	0				0	0	0	5294	9.6	42.3	3.5	92,118
0	0	32.98						0	0	1106	15.3	19.3	3.7	50,325
0	0	38.71					0	0	0	2387	14.1	32.4	2.6	55,751
0	0	55.97	0	0	0		0	0	0	2692	10.8	39.6	6.2	84,738
0	0	64.55	0	1	1		0	0		23067	12.9	45.2	3.3	76,997
0	0	62.9	0	1	1	1	0	0	0	6685	11.3	41.7	3.9	81,042
0	0	51.14	0	1	1		0	0	0	4013	4.8	57.2	3.7	131,562
0	0	57.44	0	1	1	1	0	0	0	8376	6.6	52.6	3.9	113,885
0	0	34.5		0	0			0	0	4426	14.7	25.8	4	58,855
0	0	49.27	0				0	0		11492	7	39.9	3.4	119,253
0	0	86.42						0	0	9843	16.5	39.6	4.6	95,514
0	0	57.4						0	0	9843 7590	15.3	27.2	4.7	58,013
0		57.44						0	0		6.6	52.6	3.9	113,885
	0			1						3050	18.4	22.7	5.3	56,096
	0	34.98		0			0	0	0	12733			5	82,455
0	0	71.03						0	0	2327	13.7	35.5	4.1	89,334
0		64.52	0					0	0	1315	14.9	35.9	4.1 3.2	
0 0 0	0			1	1	1	0	0	0				3.2	84,548
0 0 0	0	80.64								1710	15.1	49.9		
0 0 0 0	0	57.58	0	1	1	1	0	0	0	7506	10.9	39	3.5	86,078
0 0 0 0 0	0		0	1	1 0	1					10.9 6.1	39 48.6	3.5 2.7	103,757
0 0 0 0	0	57.58	0	1	1 0	1 D	0	0	0	7506	10.9	39	3.5	

								0 17084 1 0 7838 1 0 65410 1 0 1220 1 0 1479 1386 0 1386 1 0 1386 1 0 1386 1 0 1386 1 0 3862 1 0 3582 1 0 3582 1 0 3582 1 0 3582 1 0 3582 1 0 3582 1 0 3234 1 0 2368 1 0 2578 1 0 2578 1 0 2134 1 0 2137 1 0 3173 1 0 3187 1 0 3187 1 0 3187 1<	7084 12.1 7084 10.3 5410 14.1 1200 14.1 1409 13 6640 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1405 11.5 3092 17.3 550 7.7 4764 18.8 13173 6.6 1224 7.2 2036 17.7 2046 9 92476 13.7 2036 13.6 5719 18.8 809 12.4 2578 13.3 13175 10.4 5128 10.4 5128 13.2 13270 15.2 1475 1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 71,407 3.3 74,889 1.8 66,952 3.7 61,941 3.1 64,952 3.7 61,941 3.1 61,561 3.7 52,278 3.7 52,278 3.7 52,278 3.7 52,278 3.7 52,278 3.7 52,278 3.8 55,471 2.9 118,404 3 52,650 2.7 82,343 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 84,030 3.1 57,772 3.1 57,7732 7.5 68,693 2.8 72,129 3.1 76,633 3.1 76,633 3.2 65,489 3.3 86,469 99,8977
								0 65410 1 0 1230 1 0 15/9 4149 0 50842 1 0 1386 1386 0 1386 19651 0 4085 1 0 4085 1 0 3362 1 0 3592 1 0 3232 1 0 3232 1 0 3248 1294 0 1294 1294 0 2578 1 0 5719 2066 1294 2578 1 0 2578 1 0 2578 1 0 5128 1 0 31278 1 0 31278 1 0 3187 1 0 3278 1 0 3278 1 0 330	5410 14.1 1200 14.1 1679 13 0642 17.6 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1386 9.7 1405 11.5 3402 15.1 3173 8.6 1224 7.2 21248 9 9 9 20408 9 9 9 21244 7.2 21250 15.7 20368 17.6 527 13.7 20368 17.2 21308 14.5 1310 10.4 5128 10.4 5128 10.4 5128 13.2 1358 26.3 3357 13.3 3387 13.4 5128 13.4	$\begin{array}{c} 14.1 & 33.2 \\ 14.1 & 33.2 \\ 13 & 22.6 \\ 17.6 & 48.6 \\ 9.7 & 46.9 \\ 9.7 & 46.9 \\ 9.7 & 46.9 \\ 9.7 & 46.9 \\ 9.7 & 46.9 \\ 11.5 & 47.3 \\ 17.3 & 24.3 \\ 17.3 & 24.3 \\ 17.3 & 24.3 \\ 17.3 & 24.3 \\ 17.3 & 24.3 \\ 17.7 & 59 \\ 18.8 & 23.6 \\ 16.1 & 20.6 \\ 10.1 & 20.6 \\ 10.1$	1.8 $66,952$ 3.7 $61,941$ 3.1 $61,962$ 3.7 $61,941$ 3.1 $61,692$ 3.7 $52,278$ 3 79,609 3.7 $52,278$ 3 79,609 3.8 58,471 2.9 18,404 3 52,650 2.7 82,036 2.7 82,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.9 46,070 3.1 60,762 3.1 60,762 3.1 76,633 2.9 46,030 3.1 76,633 3.28 72,129 5 56,0403 3.5 53,170
								0 1230 1 0 16/9 0 1386 0 1386 0 1386 0 1386 0 1386 0 1386 0 1386 0 19511 0 4065 1 3562 0 4764 0 3373 0 1274 0 2778 0 2778 0 2773 0 2773 0 2773 0 2773 0 2773 0 2773 0 2773 0 2773 0 2773 0 2773 1 22689 1 3276 1 3278 1 33278 1 33278 1 33278 1	12200 (14.1 15/97) (13) 41.69 (13) 41.69 (13) 41.69 (17) 9631 (27.7 4045) (15.6 4005 (11.5 5540 (17.3 5540 (17.3 17.3 5540 (17.3 5540 (17.3 17.3 5540 (17.3 5540 (17.3 17.7 2046 (13.6 (13.6 5719 (13.6 5719 (13.6)) (13.6)) (13.6) ($\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	1.8 66,952 3.7 61,941 3.7 61,941 3.7 61,941 3.7 61,941 3.1 61,560 3.7 52,278 3.8 79,069 6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.7 7,53 3.1 57,732 3.1 57,732 2.8 50,582 2.9 64,670 3.1 76,633 2.8 50,582 2.6 71,383 4.9 99,897 3.3 55,31/00 3.5 53,120 3.5 53,190 3.4 99,7987 3.3 66,649
								0 1679 0 4149 0 50842 0 1386 0 1386 0 1386 0 1386 0 1386 0 1465 0 3592 0 3582 0 3322 0 3323 0 1234 0 2486 0 2278 1 2066 1 2063 1 2272 1 23649 1 2273 1 2273 1 22649 1 22649 1 3227 1 3327 1 33278 1 33278 1 33278 1 33278 1 33278 1 33278 1 33316 1875	16/99 13 16/99 13 08/82 17.6 13/86 9.7 13/86 9.7 13/86 9.7 13/86 9.7 13/86 9.7 13/86 9.7 13/86 9.7 5/80 7.7 4/25 15.6 3/92 16.1 3/13 8.6 2/26/2 9 2/27/2 15.1 2/26/2 13.7 2/26/8 9 9 72/2 15.7 10.6 5/719 10.8 5/80/9 12.4 2/27/2 13.7 5/27/3 10.6 5/27/3 10.6 5/27/3 10.4 5/28/2 15.2 13/27/6 10.4 5/28/2 15.2 13/27/6 13.2 13/27/6 13.2 13/28/8 26.3 <	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.7 61,941 3.7 61,941 3.1 61,640 3.7 52,278 3 79,600 4.8 45,864 1.9 64,675 3 79,636 3.8 55,671 3.8 52,660 5.1 52,660 5.1 52,660 5.1 52,660 5.1 64,675 5.2.7 82,432 2.7 82,432 2.7 82,433 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.9 46,470 2.9 46,470 2.9 46,477 5 52,453 2.8 70,212 5 53,170 3.8 56,460 3.5 33,149 4.9 99,897 3.3 86,169
								0 4149 0 50842 1 0 1386 1386 0 1386 19511 2 0 14655 1 1 0 4065 1 1 0 3532 1 1 3 0 3173 3 3322 1 3 3323 0 12744 3322 1 3 3331 3 <t< td=""><td>4169 13 6082 7.7.6 1336 9.7.7 1336 9.7.7 9951 27.7 9451 27.6 4055 15.6 4005 15.6 3330 9.7.7 3313 8.6 1224 7.2 2408 9 722 15.1 2578 16.7 2578 16.7 2720 15.1 2578 10.6 5719 18.8 899 12.4 2578 13.7 5108 17.7 2046 17.2 2173 10.6 5719 18.8 5128 10.6 5128 10.6 5128 10.0 5128 10.2 5128 10.2 5128 13.3 5138 26.3 1176 13.3 5284</td><td>$\begin{array}{cccccccccccccccccccccccccccccccccccc$</td><td>3.7 61,941 3.1 61,560 3.7 52,278 3.8 79,069 6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.3 76,227 3.1 57,732 7.5 68,693 2.9 64,720 3.1 76,732 3.1 76,033 2.8 50,582 2.9 64,720 3.5 51,100 3.5 51,210 3.5 51,210 4.9 99,897 3.3 86,169 4.9 99,897 3.3 86,169 4.9 99,897 3.3 86,169 4.9 99,897</td></t<>	4169 13 6082 7.7.6 1336 9.7.7 1336 9.7.7 9951 27.7 9451 27.6 4055 15.6 4005 15.6 3330 9.7.7 3313 8.6 1224 7.2 2408 9 722 15.1 2578 16.7 2578 16.7 2720 15.1 2578 10.6 5719 18.8 899 12.4 2578 13.7 5108 17.7 2046 17.2 2173 10.6 5719 18.8 5128 10.6 5128 10.6 5128 10.0 5128 10.2 5128 10.2 5128 13.3 5138 26.3 1176 13.3 5284	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.7 61,941 3.1 61,560 3.7 52,278 3.8 79,069 6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.3 76,227 3.1 57,732 7.5 68,693 2.9 64,720 3.1 76,732 3.1 76,033 2.8 50,582 2.9 64,720 3.5 51,100 3.5 51,210 3.5 51,210 4.9 99,897 3.3 86,169 4.9 99,897 3.3 86,169 4.9 99,897 3.3 86,169 4.9 99,897
								0 4169 0 50842 1 0 1386 1386 0 1386 19531 2 0 14625 1 1 0 4085 1 3642 1 0 3642 1 0 3540 1 0 3542 1 0 3373 0 1234 0 3224 0 2498 0 2498 0 2498 0 2578 1 0 5779 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 3277 1 0 3278 1 0 3166 1 10'05 1 0 3278 1 0 31278 1 0 33278 1 0 33278 1 0 33116 2 0 3548 2 0 3548 2 0	41269 13 6862 17.6 1336 9.7 9831 27.7 4425 15.6 4405 15.6 4405 15.6 1338 9.7 9831 27.7 4425 15.6 1338 9.7 3550 7.7 373 8.6 1224 7.2 2468 9 7272 15.1 2578 13.6 5719 18.8 9890 12.4 2578 10.6 5727 10.6 5128 10.7 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.4 5128 10.6 5128 10.2 5128 10.2 5128 13.3 5128	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.7 61,941 3.1 61,560 3.7 52,278 3.8 79,069 6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.3 76,227 3.1 57,732 7.5 68,693 2.9 64,720 3.1 76,732 3.1 76,033 2.8 50,582 2.9 64,720 3.5 51,100 3.5 51,210 3.5 51,210 4.9 99,897 3.3 86,169 4.9 99,897 3.3 86,169 4.9 99,897 3.3 86,169 4.9 99,897
								0 1386 0 1385 0 19511 2 0 14645 1 0 4065 1 0 69531 2 0 69542 1 0 5540 4764 0 33822 1 0 12744 1 0 2278 1 0 2278 1 0 5278 1 0 2738 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2174 1 0 31278 1 0 31278 1 0 31278 1 0 31278 1 0 31278 1 0 31274 1	1386 9.7 1386 9.7 9611 27.7 9621 27.7 9631 27.7 9631 15.5 9542 17.3 9540 7.7 9741 18.8 3822 16.1 1274 7.2 2496 9 9272 15.1 2578 16.7 2772 15.1 2578 16.7 20308 17.7 2046 13.6 5719 18.8 9899 12.4 2578 13.7 5108 17.2 9106 15.2 1927 10 5128 10.6 5128 10.6 5128 10.2 9735 21.4 9736 21.4 9736 21.4 9738 26.3 1176 13 1284	9.7 46.9 9.7 46.9 27.7 22 15.6 39.5 11.5 47.3 7.7 59 16.8 23.8 16.1 20.6 8.6 31 7.2 33.3 2 9 35.3 9 35.3 9 35.3 9 35.3 9 35.3 9 35.3 9 35.3 9 35.3 15.1 42 16.7 19.2 15.1 42 16.7 19.2 15.1 42 16.7 19.2 15.2 63.6 17.2 23.3 13.6 31.6 15.2 63.6 17.2 21.1 15.2 63.6 17.2 21.1 15.2 63.6 17.2 21.1 15.2 63.6 17.2 21.1 15.2 63.6 15.2 63.6 10.3 15.2 63.6 10.3 21.4 24.1 22.7 30.8 21.4 24.1 23.7 30.8 13 58 21.4 24.1 23.7 30.8 13 58 21.4 24.1 23.7 30.8 13 58 21.4 24.1 23.7 30.8 13 58 21.4 24.1 23.7 30.8 21.4 24.1 23.7 30.8 24.8 21.1 24.2 32.5 24.8 21.1 25.2 32.5 24.8 21.1 25.2 32.5 24.8 21.1 25.2 32.5 25.2 32.5 25.5 25.5 25.5 25.5 25.5 25.5 25.5	3.7 5.2278 3 79,060 6.8 45,864 1.9 61,675 3 79,358 3.8 88,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 57,732 2.9 64,720 3.1 76,833 2.8 50,582 2.8 72,129 5 82,455 2.8 50,582 2.8 72,129 5 82,455 2.4 9,9,897 3.3 5,51,100 3.5 53,120 3.5 53,120 4.9 99,897 3.3 86,169 4.9 9,8997 3.3 86,169
								0 1386 0 1385 0 19631 2 0 16425 1 0 3695 1 0 5540 3672 0 3731 3 0 1234 3133 0 1234 3273 0 2783 1 0 2036 1 0 2733 1 0 2738 1 0 2737 1 0 2737 1 0 2737 1 0 2737 1 0 2737 1 0 2578 1 0 3273 1 0 3273 1 0 3273 1 0 31872 1 0 31274 1 0 31273 1 0 31273 1	1386 9.7 1386 9.7 9631 27.7 4055 15.6 4005 15.6 4005 15.6 3922 17.3 5540 7.7 4741 18.8 3822 16.1 12124 7.2 2468 9.7 722 15.1 2578 16.7 2578 16.7 2578 16.7 2578 13.7 2046 17.2 2050 17.7 2046 17.2 2578 13.3 2579 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 527 13.3 3138 26.3 11765 10.2 5384 26.3 1274 13.3 1384	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	3.7 5.2278 3 79,060 6.8 45,864 1.9 61,675 3 79,358 3.8 88,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 57,732 2.9 64,720 3.1 76,833 2.8 50,582 2.8 72,129 5 82,455 2.8 50,582 2.8 72,129 5 82,455 2.4 9,9,897 3.3 5,51,100 3.5 53,120 3.5 53,120 4.9 99,897 3.3 86,169 4.9 9,8997 3.3 86,169
	1 0 1 1 1 0 0 0 0 0 0 1 1 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0							0 19631 2 0 14455 1 0 4065 1 0 4065 1 0 3672 1 0 3540 1 0 3132 1 0 3133 1 0 1234 1 0 2788 1 0 2578 1 0 2578 1 0 2733 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 31236 1 0 13274 1 0 31275 1 0 31274 1 0 31273 1 0 31274 1 0 31275 <t< td=""><td>96:11 27.7 64:55 15.6 40:05 11.5 36:20 17.3 55:40 7.7 37:41 18.8 3822 16.1 3133 8.6 12:24 7.2 26:30 17.7 20:46 7.2 20:40 9 72:72 15.1 25:76 13.6 7723 10.6 57:19 18.8 59:09 12.4 25:78 10.7 51:08 14.5 41:66 17.2 50:08 17.7 20:08 17.7 20:08 17.2 20:08 17.2 20:08 10.2 51:28 10.6 52:29 15.2 17:05 10.4 51:28 15.2 17:38 26.3 17:38 26.3 17:38 26.3 <t< td=""><td>$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$</td><td>6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 7,732 2.9 64,720 3.1 7,627 2.9 64,720 3.1 7,732 2.8 50,582 2.8 72,129 5 82,455 2.8 50,582 2.8 50,582 2.8 50,582 2.8 52,455 3.5 53,170 3.5 53,170 3.5 53,170 3.5 77,669 4.9 99,897 4.2 49,883 3.4 9,798 4.2 49,883 3.4 49,791 <</td></t<></td></t<>	96:11 27.7 64:55 15.6 40:05 11.5 36:20 17.3 55:40 7.7 37:41 18.8 3822 16.1 3133 8.6 12:24 7.2 26:30 17.7 20:46 7.2 20:40 9 72:72 15.1 25:76 13.6 7723 10.6 57:19 18.8 59:09 12.4 25:78 10.7 51:08 14.5 41:66 17.2 50:08 17.7 20:08 17.7 20:08 17.2 20:08 17.2 20:08 10.2 51:28 10.6 52:29 15.2 17:05 10.4 51:28 15.2 17:38 26.3 17:38 26.3 17:38 26.3 <t< td=""><td>$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$</td><td>6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 7,732 2.9 64,720 3.1 7,627 2.9 64,720 3.1 7,732 2.8 50,582 2.8 72,129 5 82,455 2.8 50,582 2.8 50,582 2.8 50,582 2.8 52,455 3.5 53,170 3.5 53,170 3.5 53,170 3.5 77,669 4.9 99,897 4.2 49,883 3.4 9,798 4.2 49,883 3.4 49,791 <</td></t<>	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	6.8 45,864 1.9 61,675 3 79,358 3.8 58,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 7,732 2.9 64,720 3.1 7,627 2.9 64,720 3.1 7,732 2.8 50,582 2.8 72,129 5 82,455 2.8 50,582 2.8 50,582 2.8 50,582 2.8 52,455 3.5 53,170 3.5 53,170 3.5 53,170 3.5 77,669 4.9 99,897 4.2 49,883 3.4 9,798 4.2 49,883 3.4 49,791 <
	1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0	1 1 0 0 0 0 0 0 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0						0 14425 1 0 4065 1 0 3562 1 0 3562 1 0 3530 1234 0 1234 1 0 2576 1 0 2576 1 0 2576 1 0 2578 1 0 2568 1 0 2578 1 0 2669 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2579 1 0 2578 1 0 2578 1 0 3527 1 0 3138 1 0 3187 1 0 3278 1 0 30116 2 0 2641 1<	Ad25 15.6 Ad25 11.5 3692 17.3 3692 17.3 3692 17.3 3747 47.64 3173 8.6 3173 8.6 3173 8.6 97272 15.1 3278 16.7 5030 17.7 5030 17.7 5030 17.7 5030 17.7 5030 17.7 5030 17.7 5030 17.7 5030 12.4 5030 12.4 5030 12.4 5030 12.4 5030 12.4 5030 12.4 5120 14.5 5120 10.4 5123 10.6 5241 11.4 8376 21.4 6016 22.7 5384 26.3 3357 9.3 3481	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	6.8 45,864 1.9 61,675 3 79,358 3.8 88,471 2.9 118,494 3 52,660 5.1 54,272 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 69,762 3.1 76,732 2.9 64,720 3.1 76,833 2.9 64,720 3.1 76,833 2.8 50,582 2.8 50,582 2.8 50,582 2.6 71,433 4.9 99,897 3.3 55,470 3.5 53,170 4.9 99,897 3.3 66,169 2.9 42,49,883 3.4 99,798 3.3 5,1310 3.4 9,9708 3.3 5,1319 3.4 49,883 </td
	1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0	1 1 0 0 0 0 0 0 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0						0 12425 1 0 4065 1 0 3562 1 0 3562 1 0 3530 1234 0 1234 1 0 2576 1 0 2576 1 0 2578 1 0 2578 1 0 2578 1 0 2579 1 0 2579 1 0 2579 1 0 2579 1 0 2579 1 0 2579 1 0 2579 1 0 2578 1 0 3527 1 0 3128 1 0 3238 1 0 3257 1 0 3678 2 0 30116 2 0 2688 13	Ad25 15.6 Ad25 11.5 3492 17.3 3492 17.3 3492 17.3 3492 18.8 3173 8.6 1244 7.2 1244 7.2 1246 7.2 2707 15.1 5038 17.7 5039 12.4 606 9 9 9 7272 15.1 5038 17.7 5039 12.4 5046 13.6 7272 11.3 51308 14.5 51308 14.5 51308 14.5 51308 14.5 51308 10.4 5132 11.4 8376 21.4 0116 23.7 5384 26.3 3387 9.3 3381 26.4 2940 11.9 2940	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	1.9 61,675 3 79,356 3.8 58,471 2.9 118,494 3 52,650 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.3 16,277 3.1 59,7732 3.1 50,552 2.8 72,129 5 50,552 2.6 71,833 3.5 33,170 3.5 33,170 3.5 53,170 3.5 77,359 3.4 99,9997 4.2 49,883 3.4 99,708 4.2 49,883 3.4 49,71 3.4 49,471
	1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 0 0 0 1 1 0						0 4065 1 0 3652 1 0 4764 1 0 33822 1 0 3173 1 0 12744 1 0 2248 1 0 2273 1 0 5278 1 0 2269 1 0 2273 1 0 2773 1 0 2773 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 3274 1 0 31268 1 0 13278 1 0 32278 1 0 32378 1 0 3364 2 0 3578 2 0 36678	4065 11.5 3962 17.3 5540 7.7 3744 18.8 3822 16.1 3133 8.6 1224 7.2 2498 9 2272 15.1 2578 16.7 2030 17.7 2046 13.6 5719 18.8 8999 12.4 2578 13.7 5108 17.7 2046 17.2 2173 10.6 5719 18.8 8999 12.4 2578 13.7 5128 10.6 5128 10.6 5128 10.6 5128 10.6 526 15.2 7975 21.4 0116 23.7 3384 26.3 12766 13 31381 26.4 2940 11.9 3150	$ \begin{aligned} &11.5 & 47.3 \\ &17.3 & 24.3 \\ &17.7 & 59 \\ &18.8 & 23.6 \\ &16.1 & 20.6 \\ &8.6 & 31 \\ &7.2 & 33.3 & 2 \\ &9 & 35.3 \\ &9 & 35.3 \\ &9 & 35.3 \\ &9 & 35.3 \\ &9 & 35.3 \\ &15.1 & 42 \\ &16.7 & 19.2 \\ &16.7 & 19.2 \\ &16.7 & 19.2 \\ &17.7 & 23.9 \\ &13.6 & 31.6 \\ &16.2 & 33.3 \\ &15.1 & 42 \\ &16.7 & 19.2 \\ &17.7 & 23.9 \\ &13.6 & 31.6 \\ &17.7 & 23.9 \\ &13.6 & 31.6 \\ &17.7 & 23.9 \\ &13.6 & 34.4 \\ &13.7 & 35.5 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &14.5 & 23.6 \\ &15.2 & 63.6 \\ &17.2 & 21.1 \\ &13.6 \\ &15.2 & 63.6 \\ &14.4 & 24.1 \\ &23.7 & 30.8 \\ &21.4 & 24.1 \\ &23.7 & 30.8 \\ &21.4 & 24.1 \\ &23.7 & 30.8 \\ &21.3 & 58 \\ &10 & 33.7 \\ &9.3 & 24.8 \\ &19.9 \end{aligned} $	3 79,356 3.8 58,471 2.9 118,494 3 52,650 5.1 54,272 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.3 76,227 3.1 69,762 3.1 57,732 2.9 64,720 3.1 76,633 2.8 50,582 2.8 72,120 5 82,455 2.6 71,433 4.9 99,897 3.3 56,469 3.5 53,170 4.9 99,897 3.3 86,164 4.9 99,897 3.3 86,164 4.9 99,897 3.3 86,164 4.2 49,883 3.4 99,708 4.2 49,883 3.4 49,718 3.4 49,471
	1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 0 0 0 1 1 0						0 3692 1 0 5540 0 4764 1 0 3173 1 0 1294 1 0 1294 1 0 2576 1 0 2576 1 0 2576 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 3527 1 0 3184 1 0 3128 1 0 3278 1 0 3278 1 0 4378 2 0 3548 2	3692 17.3 3692 17.3 4754 18.8 3173 8.6 3173 8.6 3173 8.6 3173 8.6 3173 8.6 3173 8.6 3173 8.6 3173 8.6 3173 15.1 5038 17.7 5038 17.7 5038 17.7 5038 12.4 5059 12.4 5027 1.3 5277 1.3 5279 1.3 5279 1.3 5126 14.5 5126 10.6 5127 10.6 5128 10.6 529 12.4 5126 13 528 26.3 1358 26.3 1358 26.3 1351 10.1 512 1.4 513 10.1	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	3.8 58,471 2.9 118,494 3 52,650 5.1 54,272 2.7 82,434 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 3.1 69,762 3.1 57,732 3.1 57,732 3.1 56,652 2.8 72,129 5 82,6455 2.6 71,833 4.9 56,469 3.5 53,170 3.5 53,170 3.5 53,170 3.4 69,087 4.9 9,0807 4.2 49,863 3.4 89,169 3.4 89,708 3.4 49,471
	1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 0 0 0 1 1 0						0 5540 0 4764 1 0 33822 1 0 3173 1 0 12744 1 0 2278 1 0 2278 1 0 2578 1 0 5038 1 0 2046 1 0 2038 1 0 2038 1 0 2038 1 0 2039 1 0 2036 1 0 2036 1 0 2036 1 0 2578 1 0 2578 1 0 31274 1 0 31274 1 0 31278 1 0 31278 1 0 31278 1 0 31278 1 0 31876 2	5540 7.7 744 18.8 3822 16.1 3832 16.1 3133 8.6 1224 7.2 2470 9 2721 15.1 2578 16.7 2038 17.7 2046 13.6 5719 18.8 5899 12.4 2578 13.7 5108 17.2 10.6 5719 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 10.6 5128 13.2 5241 11.4 8378 23.3 1726 13.3 1238 26.3 12748 13.2 1281 26.4 2940 11.9 1350	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	$\begin{array}{cccc} 2.9 & 118,494 \\ 3 & 52,650 \\ 5.1 & 54,272 \\ 2.7 & 82,344 \\ 2.7 & 83,036 \\ 2.3 & 76,227 \\ 3.1 & 97,752 \\ 3.1 & 97,752 \\ 7.5 & 68,693 \\ 2.9 & 64,720 \\ 3.1 & 76,833 \\ 2.8 & 50,582 \\ 2.8 & 72,129 \\ 5 & 82,455 \\ 2.8 & 72,129 \\ 5 & 82,455 \\ 3.5 & 51,310 \\ 4.9 & 99,897 \\ 3.8 & 56,489 \\ 3.5 & 53,130 \\ 4.9 & 99,897 \\ 3.3 & 86,169 \\ 4.9 & 99,897 \\ 4.2 & 49,883 \\ 3.4 & 89,708 \\ 3.4 & 49,711 \\ 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 44,677 \\ 3.5 & 3.5 & 3.5 & 3.5 \\ 3.5 & 3.5 & 3.5 & 3.5 & 3.5 \\ 3.5 & 3.5 & 3.5 & 3.5 & 3.5 & 3.5 \\ 3.5 & 3.5 & 3.5 & 3.5 & 3.5 & 3.5 & 3.5 \\ 3.5 & 3$
								0 4764 1 0 3822 1 0 1234 1234 0 1234 1234 0 2268 1234 0 2576 1 0 2576 1 0 2576 1 0 2579 1 0 2579 1 0 2579 1 0 2579 1 0 2579 1 0 2577 1 0 3174 1 0 3168 1 0 3187 1 0 31274 1 0 31273 1 0 31276 1 0 31278 1 0 32378 1 0 32578 1 0 33578 2 0 30116 2 0 2648	4764 18.8 4764 18.8 3173 8.6 3173 8.6 3173 8.6 3173 8.6 3173 8.6 9 9 7272 15.1 5038 17.7 5038 17.7 5036 13.6 2723 10.6 5079 18.8 5699 12.4 5120 11.3 5277 1.3 5276 10.6 5128 10.6 5126 14.2 5127 10.4 5128 10.4 5128 10.4 5128 13.2 7035 21.4 0116 23.7 538 26.3 13387 9.3 1381 26.4 2940 11.9 1300 10.9	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	3 52,650 5.1 54,272 2.7 82,336 2.7 83,036 2.7 83,036 2.7 83,036 2.7 83,036 2.3 16,227 3.1 97,732 3.1 97,732 2.0 64,770 3.1 76,633 2.8 72,129 5 82,455 2.6 71,833 4.9 56,469 3.5 53,110 3.5 53,210 4.1 67,133 3.5 53,210 4.2 49,863 4.3 9,9097 3.3 86,169 4.2 49,863 3.4 99,708 3.4 99,709 3.4 49,474
	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 1 1 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0						0 33622 1 0 3173 0 12244 0 2246 0 2272 0 5278 0 2578 0 2046 0 2773 0 5739 0 5779 0 2578 0 2578 0 2578 0 2578 0 2578 0 2578 0 2578 0 2578 0 2578 0 3274 0 31274 0 31274 0 31274 0 31278 0 31278 0 31278 0 31278 0 3167 0 30116 2 3548 0 2668 0 3357 18678	3822 10.1 3823 10.1 1294 7.2 1294 7.2 2470 9 12972 15.1 2578 10.7 2040 13.6 5719 18.8 9899 12.4 2578 13.7 3527 11.3 2174 15.2 1508 17.2 1006 17.2 1015 10.4 5128 10.6 527 11.3 2174 15.2 1055 10.4 5128 10.6 527 14.5 4164 17.2 7975 10.4 5128 10.6 524 15.2 1354 26.3 11766 13 3387 26.3 1274 13.2 128.1 26.4 2940 11.9 1350	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
								0 3173 0 1294 0 2498 0 2772 1 0 2578 1 0 2578 1 0 2578 1 0 2064 1 0 2773 1 0 2064 1 0 2773 1 0 2773 1 0 2773 1 0 2773 1 0 2579 1 0 2576 1 0 25662 1 0 25662 1 0 25662 1 0 25662 1 0 25662 1 0 25662 1 0 2568	3173 8.6 3173 8.6 1244 7.2 1247 7.2 1248 9 9 9 7272 15.1 5038 17.7 5038 17.7 5038 17.7 5039 12.4 5070 13.8 5079 13.8 5079 14.8 5108 14.5 5105 10.4 5126 10.6 5127 10.6 5128 10.6 5129 11.4 8376 21.4 0116 23.7 5188 26.3 11786 13 5181 20.4 5182 13 5182 13 5182 23.3 13357 9.3 1381 26.4 2940 11.9	8.6 31 7.2 33.3 9 35.3 15.1 42 9 35.3 15.1 42 17.7 23.9 16.7 19.2 17.7 23.9 13.6 31.6 10.6 41.7 13.6 31.6 11.3 40 15.2 63.6 17.2 21.1 10.4 28.3 10.6 28.2 10 32.5 11.3 40 15.2 63.6 10 32.3 10.6 28.2 10.3 56 11.4 34.7 13 56 11.4 34.7 13 56 11.3 58 21.4 24.1 23.7 30.8 24.3 58.4 10 33.7 9.3 24.8	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
	0 0 0 0 0 0 0 1 1 1 0 0 0 0 0 1 1 1 1 0							0 1244 0 1244 0 2246 0 2273 0 5038 0 2046 0 2737 0 5739 0 5773 0 5773 0 2578 1 2578 0 2578 1 2578 0 2578 1 3277 0 2174 1 1005 1 10705 1 10705 1 31278 1 31278 1 31278 1 25662 1 26641 0 43378 0 3116 2 3548 0 3357 1 3357 1 3357 1 3357 1 3357 1 3357 </td <td>1244 7.2 1244 7.2 1244 7.2 1245 9 1272 15.1 12578 16.7 12578 13.6 5719 18.8 9899 12.4 2578 13.7 5108 17.2 13508 14.5 1426 17.2 1705 10.4 5128 10.6 5279 21.4 9785 21.4 9878 22.3 1352 21.4 9878 23.3 1358 26.3 11726 13.3 13387 23.3 1384 26.3 1326 10.0 3337 9.3 1381 26.4 2400 11.9</td> <td>$\begin{array}{cccccccccccccccccccccccccccccccccccc$</td> <td>$\begin{array}{cccccccccccccccccccccccccccccccccccc$</td>	1244 7.2 1244 7.2 1244 7.2 1245 9 1272 15.1 12578 16.7 12578 13.6 5719 18.8 9899 12.4 2578 13.7 5108 17.2 13508 14.5 1426 17.2 1705 10.4 5128 10.6 5279 21.4 9785 21.4 9878 22.3 1352 21.4 9878 23.3 1358 26.3 11726 13.3 13387 23.3 1384 26.3 1326 10.0 3337 9.3 1381 26.4 2400 11.9	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
	0 1 0 0 1 1 0 1 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0							0 1294 0 2496 0 2772 1 0 2576 1 0 2576 1 0 2056 1 0 2026 1 0 2773 1 0 5719 1 0 22699 1 0 3527 1 0 3528 1 0 1765 1 0 3528 1 0 3558 1	1244 7.2 1246 9 7272 15.1 5278 16.7 5018 17.7 5026 13.6 2727 10.6 5759 16.8 5699 12.4 5579 16.2 5579 16.2 5570 16.2 5106 14.5 5126 10.6 5127 10.4 5128 10.6 5128 10.4 5128 10.4 5128 13.2 6070 21.4 0116 23.7 5382 26.3 13582 26.3 3357 9.3 1381 20.4 2940 11.9	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	2.7 83,036 2.3 76,227 3.1 99,762 7.5 68,693 2.9 64,730 3.1 76,633 2.9 64,730 3.1 76,633 2.8 50,582 2.8 72,129 5 82,2455 2.6 71,833 4.9 99,897 3.8 55,489 3.5 54,89 3.5 54,89 3.5 54,89 3.5 54,89 3.5 64,80 4.9 99,897 3.3 86,169 4.9 99,897 4.2 49,883 3.4 89,708 3.3 49,719 3.3 44,77
	1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 0 0 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0						0 2468 0 7272 1 0 2578 1 0 5038 1 0 2036 1 0 2737 1 0 5739 1 0 57739 1 0 2578 1 0 2578 1 0 2578 1 0 2174 1 0 5128 1 0 12757 1 0 31274 1 0 31278 1 0 32278 1 0 32737 1 0 32737 1 0 32738 1 0 3374 1 0 33014 2 0 3364 2 0 3357 2 0 3357 1 0 1386 1	9 2498 9 2498 9 272 25 15 1 2578 16 7 272 24 15 1 2578 10 6 5719 18 8 99 12 4 2578 13 7 2174 15 2 1705 10 4 5128 10 6 15 2 1875 10 5128 175 10 5128 175 10 5128 175 2 14 4 5128 175 1 52 1875 1 52 1875 1 5 2 1875 1 5 2 1875 1 5 2 1 1 5 1 1 5 1 1 1 1 1 1 1 1 1 1	9 35.3 15.1 42 16.7 192 17.7 23.9 13.6 31.6 13.6 31.6 13.7 35.5 13.4 38.4 13.7 35.5 13.4 40 15.2 63.8 17.2 21.1 13.4 40 15.2 63.8 10.4 22.3 10.6 29.2 11.4 38.7 15.2 63.8 21.4 24.1 23.7 30.8 21.4 24.1 23.7 30.8 13.8 58 14.3 58 14.3 58 14.3 58 14.3 58 14.3 58 14.3 58 15.2 63.8 15.2 63.8 25.4 24.1 25.7 30.8 25.8 27.3 25.8 27.5 25.8 27.5 25.8 27.5 25.8 27.5 25.9 27.5 25	$\begin{array}{cccc} 2.3 & 76,227\\ 3.1 & 69,752\\ 3.1 & 57,732\\ 7.5 & 68,693\\ 2.9 & 64,720\\ 3.1 & 76,633\\ 2.8 & 50,582\\ 2.8 & 72,129\\ 5 & 82,455\\ 5 & 82,455\\ 2.6 & 71,433\\ 4.9 & 99,897\\ 3.8 & 56,489\\ 3.5 & 53,170\\ 4.1 & 67,135\\ 3.5 & 77,669\\ 4.9 & 99,897\\ 3.3 & 86,169\\ 4.9 & 99,897\\ 4.2 & 49,883\\ 3.4 & 89,798\\ 3.4 & 89,798\\ 3.4 & 49,471\\ 3.4 & 49,471\\ 3.5 & 44,677\\ \end{array}$
	0 0 1 0 0 1 1 0 1 1 0 0 0 0 0 0 0 0 0 0							0 7222 1 0 2576 1 0 2508 1 0 2723 1 0 2773 1 0 2773 1 0 2773 1 0 2578 1 0 2578 1 0 2577 1 0 2174 1 0 3168 1 0 5128 1 0 5128 1 0 3273 1 0 3273 1 0 3278 1 0 3278 1 0 3278 1 0 25662 1 0 25662 1 0 4378 2 0 30116 2 0 2688 3357 0 3357 1 0 2940 1	7272 15.1 7278 16.7 5038 17.7 5038 17.7 5038 17.7 5039 13.6 2723 10.6 5759 16.8 5690 12.4 5570 11.3 5527 11.3 51306 14.5 51308 14.6 5127 15.2 1318 10.6 5128 10.4 5128 10.4 5128 13.2 6562 15.2 7035 21.4 011.6 23.7 5382 26.3 1726 13 5382 26.3 3357 9.3 152.1 15.2 138.1 26.4 24.2 10 3357 9.3 152.1 26.2 1300 10.9	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	3.1 09,762 3.1 19,732 3.1 19,732 3.1 17,732 3.1 176,633 2.0 04,770 3.1 76,633 2.8 172,120 5 82,645 2.6 17,183 4.9 56,469 3.5 53,110 3.5 53,210 4.1 67,135 3.5 77,369 4.9 90,907 3.3 86,169 4.9 90,907 4.2 49,863 3.4 99,708 3.4 90,708 3.4 49,471 4.2 49,863 3.4 49,471 4.5 44,677
	0 1 0 0 1 1 1 0 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0				0 0 0 0		0 2578 1 0 5038 1 0 2046 1 0 2731 1 0 2773 1 0 2578 1 0 2578 1 0 2578 1 0 2578 1 0 2174 1 0 3108 1 0 12705 1 0 31278 1 0 31278 1 0 32278 1 0 32737 1 0 32737 1 0 32738 1 0 3367 2 0 30116 2 0 2668 1 0 3357 1 0 3357 2 0 2668 1 0 1360 1	2278 16.7 2026 13.6 2030 17.7 2046 13.6 5719 13.6 5719 18.8 8999 12.4 2578 13.7 3527 11.3 2124 15.2 1508 14.5 4126 17.2 1705 10.4 5128 10.6 5278 13.7 562 15.2 7935 21.4 0014 23.7 3584 26.3 11726 13.3 1284 10.0 3387 9.3 1381 26.4 2940 11.9	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
	1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0							0 5038 1 0 2046 1 0 2723 1 0 5719 1 0 2559 1 0 2576 1 0 2577 1 0 2576 1 0 2577 1 0 2576 1 0 2577 1 0 35277 1 0 31278 1 0 31278 1 0 31278 1 0 33278 1 0 33578 1 0 3357 1 0 357 1 0 357 1 0 357 1 0 357 1 0 357	5038 17.7 2046 13.6 2723 10.6 5719 18.8 5699 12.4 5276 13.7 3527 11.3 3527 11.3 3527 15.2 5108 10.6 5128 10.6 5127 15.2 7035 21.4 6562 15.2 7035 21.4 6011.6 23.7 5388 26.3 11786 10 3387 9.3 31851 26.4 1380 10.9	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
	0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 1 0 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 0			0 0 0 0 0 0 0 0 1	0 0 0 0 0 0		0 2046 1 2723 1 0 2759 1 0 26899 1 0 2578 1 0 3527 1 0 31274 1 0 31274 1 0 31274 1 0 31278 1 0 31265 1 0 2565 1 0 2565 1 0 2668 1 0 30116 2 0 3578 2 0 3588 2 0 3116 1 0 2668 1 0 2668 1 0 1367 1 0 1360 1 1 0 1360 1 1 0 1360 1 0 10 1 0 1	2046 13.6 2723 10.6 5719 18.8 5999 12.4 2578 13.7 3527 11.3 2174 15.2 2578 13.7 31.7 31.7 31.7 31.7 31.7 31.7 31.7	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccc} 2.9 & 4.720 \\ 3.1 & 76,633 \\ 2.8 & 50,582 \\ 2.8 & 72,129 \\ 5 & 82,455 \\ 2.6 & 71,233 \\ 4.9 & 99,897 \\ 3.8 & 56,489 \\ 3.5 & 53,170 \\ 4.1 & 67,135 \\ 3.5 & 77,669 \\ 4.9 & 99,897 \\ 3.3 & 86,169 \\ 4.9 & 99,897 \\ 4.2 & 49,883 \\ 3.4 & 89,798 \\ 3.4 & 49,713 \\$
	0 0 1 0 1 0 0 0 0 1 1 0 0 0 0 0 0 0 0 0	0 1 1 0 1 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0						0 2723 1 0 5719 1 0 2589 1 0 2578 1 0 3577 1 0 3571 1 0 3106 1 0 1705 1 0 31278 1 0 32778 1 0 32778 1 0 32778 1 0 32778 1 0 32778 1 0 32778 1 0 32662 1 0 25662 1 0 3678 2 0 30116 2 0 2688 2 0 3357 1 0 3357 1 0 2688 1 0 1360 1	2723 10.6 5719 16.8 8699 12.4 5276 13.7 3527 11.3 3527 11.3 3527 11.3 3527 11.3 3527 11.3 3527 10.6 5128 10.6 5127 10.2 152 10.6 5127 10.2 166 11.4 8376 21.4 0116 22.7 5542 26.3 11726 13 1387 9.3 3387 9.3 1381 26.4 2940 11.9	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.1 76.633 2.8 50.582 2.8 72.129 5 82.455 2.6 71.833 3.8 56.489 3.5 53.170 4.1 67.135 4.5 99.897 3.8 56.489 4.3 67.135 4.3 99.897 4.3 99.897 4.3 99.897 4.3 99.893 2.4 49.863 2.9 83.708 3.4 99.708 3.4 99.708 3.4 49.471 4.2 49.863 3.4 49.471 4.5 44.677
	1 0 1 0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0	1 0 1 0 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0 0				0 0 0 0		0 573.9 1 0 25578 1 0 3527.1 1 0 2174.1 1 0 5128 1 0 1705.1 1 0 31278.1 1 0 31278.1 1 0 31278.1 1 0 25652.1 1 0 25662.1 1 0 46378 2 0 30116.2 2 0 3587 1 0 35377 1 0 3548.2 2 0 3548.2 2 0 3548.2 2 0 2668 1861.2 0 1360.1 1	5719 18.8 6999 12.4 2578 13.7 3527 11.3 2174 15.2 5108 14.5 4166 17.2 1705 10.4 5128 10.6 5127 11.3 2174 15.2 1875 10 5542 15.2 7935 21.4 8078 21.4 8078 21.4 8078 21.4 9034 23.7 1384 26.3 11726 13 3387 9.3 1381 26.4 2420 11.9 2940 11.9	18.8 21.1 12.4 38.4 13.7 35.5 11.3 40 15.2 63.6 17.2 21.1 14.5 29.6 17.2 21.1 10.6 29.2 10.4 28.3 10.6 29.2 11.4 38.7 15.2 63.6 21.4 24.1 23.7 30.6 24.4 24.1 23.7 30.8 13 58 10 33.7 9.3 24.8 10.3 33.7	$\begin{array}{cccc} 2.8 & 50.582 \\ 2.8 & 72.129 \\ 5 & 82.455 \\ 2.6 & 71.635 \\ 3.4 & 9 & 99.897 \\ 3.8 & 56.489 \\ 3.5 & 53.170 \\ 4.1 & 67.135 \\ 3.5 & 77.669 \\ 4.9 & 99.897 \\ 3.3 & 86.169 \\ 4.9 & 99.897 \\ 4.2 & 49.883 \\ 2.9 & 83.918 \\ 3.4 & 89.798 \\ 3.4 & 49.471 \\ 3.4 & 49.471 \\ 3.4 & 49.471 \\ 3.4 & 44.477 \end{array}$
	0 1 0 1 1 0 0 0 1 1 1 1 0 0 0 0 0 0 0 0							0 5739 1 0 26899 1 0 3578 1 0 3257 1 0 2174 1 0 2174 1 0 3108 1 0 1205 1 0 31278 1 0 3278 1 0 25662 1 0 25652 1 0 43378 2 0 30146 2 0 30156 2 0 33648 2 0 3357 3357 0 13857 2688 0 2940 1 0 1340 1	5719 18.8 9899 12.4 2578 13.7 3527 11.3 2174 15.2 5108 14.5 4166 17.2 1705 10.4 5128 10.6 5127 10.2 562 15.2 7795 21.4 8078 21.4 0116 23.7 3584 26.3 11786 13 3387 9.3 1381 264 2940 11.9 2940 10.9	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	2.8 72,120 5 82,6455 2.6 71,833 4.9 99,897 3.8 55,439 3.5 51,2170 3.4 67,1355 4.1 67,1355 3.5 77,969 4.9 99,897 4.2 49,863 2.9 83,748 3.4 89,708 3.4 89,708 3.4 49,883 3.4 49,471 4.5 44,677
	1 0 1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0							0 26899 1 0 2578 1 0 3527 1 0 3127 1 0 3108 1 0 1166 1 0 11205 1 0 31279 1 0 33279 1 0 25662 1 0 25662 1 0 25662 1 0 43378 2 0 30116 2 0 3584 2 0 3584 2 0 3587 2 0 3587 2 0 3587 2 0 2688 2 0 1360 1 0 1360 1	9899 12.4 9278 13.7 3527 11.3 3527 11.3 3527 11.3 3527 11.3 3527 11.3 3527 11.6 2174 15.2 100 15.2 100 15.2 100 15.2 101 15.6 2041 11.4 8376 21.4 101.6 23.7 55.8 26.3 11726 13 3387 9.3 1381 26.4 24.9 10 3387 9.3 1851 26.4 2940 11.9	$\begin{array}{rrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrrr$	2.8 72,129 5 82,645 2.6 71,833 4.9 99,897 3.8 56,489 3.5 53,170 4.1 67,135 4.9 99,897 3.3 86,169 4.9 99,897 4.3 99,897 4.4 99,897 4.2 49,883 2.9 83,918 3.4 99,796 3.4 49,471 4.5 44,677
	0 1 0 0 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0							0 2578 1 3527 1 0 3227 1 0 106 1 0 106 1 0 4166 1 1005 1 0 5128 1 0 32278 1 0 32278 1 0 25662 1 0 7935 2 12641 1 0 46378 2 0 30116 2 0 3548 2 0 3116 2 0 3558 2 0 3357 2 0 3558 2 0 3357 1 0 1360 1 0 1360 1	2578 13.7 3527 11.3 2124 15.2 5108 14.5 4166 17.2 1705 10.4 5128 10.6 5127 10 5128 10.6 527935 21.4 8078 21.4 8078 21.4 9014 23.7 3584 26.3 11786 13.3 12841 10.1 3384 26.3 11284 20.4 11284 26.4 11384 26.3 11294 13.3 11204 24.4 1130 10.9	13.7 35.5 11.3 40 152.63.6 63.6 172.21.1 63.6 10.622.21.1 63.6 10.622.2 63.6 10.52.63.6 63.6 11.438.7 13.5 21.42.4.1 23.7 22.63.2 21.3 13.58 58.1 10.33.7 9.3 9.32.428.4 19.9	5 82,455 2.6 71,033 4.9 99,897 3.8 56,489 3.5 53,170 4.1 67,135 3.5 77,069 4.9 99,897 3.3 86,169 4.9 99,897 4.2 49,883 2.9 83,918 3.4 89,708 4.2 49,883 3.4 49,471 4.5 44,677
	0 1 0 0 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0							0 3527 1 0 2174 1 0 5108 1 0 1166 1 0 1205 1 0 31278 1 0 32378 1 0 25652 1 0 25652 1 0 25641 1 0 48378 2 0 3156 2 0 3558 2 0 3588 2 0 3588 2 0 3588 2 0 3581 2 0 2648 1 0 2948 1 0 1360 1	3527 11.3 2174 15.2 5108 14.5 1456 17.2 1705 10.4 5128 10.6 3279 15.2 214 15.2 3279 15.2 2041 11.4 8376 2.1.4 0216 2.3 5586 26.3 11766 13 3387 9.3 1851 26.4 1.4 13.3 1126 10 3387 9.3 1126 26.3 11276 13 3387 9.3 1851 26.4 2940 11.9	11.3 40 15.2 6.3.6 14.5 29.6 17.2 21.1 10.4 28.3 10.6 29.2 15.2 63.6 21.4 28.1 10 33.2 15.2 63.6 21.4 24.1 23.7 30.8 21.4 24.1 23.7 30.8 13 56 21.3 13 13 8 10 33.7 9.3 24.8 19.9 26.4	2.6 71,833 4.9 99,897 3.8 56,489 3.5 53,170 4.1 67,135 3.5 77,969 4.9 99,897 3.3 86,160 4.9 99,897 4.2 49,883 2.9 83,918 3.4 99,796 4.2 49,863 3.4 49,471 4.5 44,677
	1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0							0 2174 1 0 5108 1 0 4166 1 1 705 1 0 31278 1 0 33278 1 0 25662 1 0 7935 2 1 2641 1 0 46378 0 30116 2 0 3548 2 0 3548 2 0 3116 2 0 3548 2 0 3548 2 0 3557 1 0 2668 1 0 2668 1 0 1360 1 0 1360 1	2174 152 5108 14.5 5108 14.5 5108 14.5 5108 17.2 5105 10.4 5128 10.6 5128 10.6 552 15.2 7795 21.4 552 21.4 552 21.4 5024 11.4 8378 23.8 733 548 26.3 11786 13 3387 9.3 3387 9.3 1381 26.4 2940 11.9 1300 10.9	152 63.6 14.5 28.6 17.2 21.1 10.4 28.3 10.6 29.2 15.2 63.6 15.4 24.1 15.2 63.6 21.4 24.1 23.7 30.8 21.4 24.1 23.7 30.8 10 33.7 9.3 58.4 10 33.7 9.3 24.8 9.3 24.8 28.4 19.9	4.9 99,897 3.8 56,469 3.5 53,170 4.1 67,135 3.5 77,969 4.9 99,897 3.3 86,169 4.9 99,897 4.2 49,883 2.9 83,918 3.4 89,798 4.2 49,883 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0							0 5108 1 0 4146 1 0 5128 1 0 5128 1 0 33270 1 0 25662 1 0 7935 1 0 25662 1 0 7935 2 0 25662 1 0 3016 2 0 3016 2 0 3548 2 0 3548 2 0 41786 0 2688 0 3357 1861 2 0 2940 1 0 1360 1	5108 14.5 1456 17.2 1705 10.4 1512 10.6 1279 15.2 150 10 15662 15.2 2010 15.2 2021 11.4 8378 21.4 1014 23.7 5582 26.3 11726 13 2028 10 3337 9.3 1851 26.4 11.9 13.3 1261 11.4 13387 9.3 1337 9.3 1351 26.4 1300 10.9	14.5 29.6 17.2 21.1 104 28.3 105 28.2 152 63.6 21.4 28.1 152 63.6 21.4 24.1 23.7 30.8 21.4 24.1 23.7 30.8 13 56 21.3 13 13 58 10 33.7 9.3 24.8 9.3 24.8 26.4 19.9	3.8 56,489 3.5 53,170 4.1 67,135 3.5 77,969 4.9 88,169 4.9 99,897 4.2 49,883 2.9 83,748 3.4 69,798 3.4 49,863 3.4 49,863 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0				0 0 0 1 0 0 1 0 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 0 1 0 <	0 0		0 4166 1 0 1705 1 0 33278 1 0 33278 1 0 25662 1 0 7935 2 1 2641 1 0 46378 0 8678 2 0 30116 2 0 3548 2 0 3116 2 0 3548 2 0 3548 2 0 3557 1 1 2668 1 0 2668 1 0 1360 1 1 357 1 0 1360 1	4166 17.2 1705 10.4 5128 10.6 5127 15.2 1875 10 5662 15.2 1875 10 5662 15.2 1875 10 5662 15.2 1878 13 6078 21.4 0116 23.7 3548 26.3 11766 10 3357 9.3 1861 26.4 2940 11.9 1300 10.9	17.2 21.1 10.4 28.3 10.6 29.2 15.2 63.6 10 33.2 15.2 63.6 21.4 24.1 23.7 30.8 21.4 24.1 23.7 30.8 10 33.7 9.3 58.4 10 33.7 9.3 24.8 9.3 24.8	3.5 5.3,170 4.1 67,135 3.5 77,969 4.9 99,897 3.3 86,169 4.9 99,897 4.2 49,883 2.9 83,918 3.4 89,798 4.2 49,883 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 1 1 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0					0 1705 1 0 5128 1 0 33278 1 1 1875 0 25642 1 0 7955 2 0 12641 1 0 8678 2 0 30116 2 0 3548 2 0 3116 2 0 3548 2 0 41786 0 2668 0 3357 1 1851 2 0 1360 1	1705 10.4 1512 10.6 3278 15.2 1875 10 5562 15.2 17975 21.4 8078 21.4 8078 21.4 8078 21.4 8078 21.4 8078 20.3 3548 26.3 1726 13 2688 10 3357 9.3 1861 26.4 2940 11.9 3360 10.9	104 28.3 106 29.2 152 63.6 10 33.2 15 63.6 214 24.1 11.4 38.7 21.4 24.1 21.4 24.1 23.7 30.8 13 56 21.3 13 13 58 10 33.7 9.3 24.8 19.9 26.4	4.1 67.355 3.5 77.969 4.9 99.897 3.3 86.160 4.9 49.893 2.9 83.918 3.4 69.796 4.2 49.863 3.4 49.471 4.5 44.677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 1 0 0 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1		1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 5128 1 0 33278 1 0 1875 0 25662 1 0 7935 2 1 2641 1 0 8678 2 0 30146 2 0 30146 2 0 3548 2 0 41786 0 2688 0 3357 0 1861 2 0 2940 1 0 1360 1	5128 10.6 3278 15.2 1877 10 5662 15.2 7935 21.4 2641 11.4 3878 21.4 0116 23.7 3548 26.3 1015 23.7 9548 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	106 29.2 15.2 63.6 10 33.2 15.2 65.6 21.4 24.1 13 58 21.4 24.1 23.7 30.8 23.7 30.8 23.3 30.8 10 33.7 9.3 24.8 26.4 19.9	3.5 77,969 4.9 99,897 3.3 86,169 4.9 99,897 4.2 49,883 2.9 83,918 3.4 89,796 3.4 49,471 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 1 0 0 0 0 0 0 0 0 0 0 1 0	0 1 0 0 0 0 0 0 0 0 1	0 1 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1		1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0			0 33278 1 1 1875 0 25642 1 0 7935 2 2 0 12641 1 0 8678 2 0 30116 2 0 3548 2 0 41786 0 2688 0 3357 1851 2 0 1360 1	15.2 12775 10 55662 15.2 7935 21.4 2641 11.4 8378 8078 21.4 0116 23.7 3548 26.3 1786 3357 9.3 1861 26.4 11.9 1360	15.2 63.6 10 33.2 5 15.2 63.6 - 21.4 24.1 - 13 56 - 21.4 24.1 - 13 56 - 21.4 24.1 - 23.7 30.8 - 26.3 21.3 - 10 33.7 - 9.3 24.8 - 26.4 19.9 -	4,9 99,897 3.3 86,169 4,9 99,867 4,2 49,863 2,9 83,918 3,4 69,796 4,2 49,863 3,4 69,796 4,2 49,863 3,4 49,871 3,4 49,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 1 0 0 0 0 0 0 0 0 0 0 1 0	0 1 0 0 0 0 0 0 0 0 1	0 1 0 0 0 0 0 1 1 1 1 1 1 1 1 1 1		0 0 0 0 1 0 0 0 0 0 0 1 0 0 0 0			0 1875 0 25662 1 0 7935 2 1 2641 1 0 46378 0 8678 2 0 30116 2 0 3548 2 0 41786 0 2688 0 3357 0 1861 2 0 1360 1 0 1360 1	1275 10 55662 15.2 79735 21.4 2641 11.4 8378 13 8678 21.4 10116 23.7 3548 26.3 11786 13 3557 9.3 18641 26.4 2940 11.9 1360 10.9	10 33.2 3 15.2 63.6 21.4 24.1 11.4 38.7 3 3 21.4 24.1 23.7 30.8 23.7 30.8 3 21.3 13 58 10 33.7 10 33.7 9.3 24.8 26.4 19.9 39.4	3.3 86,169 4.9 99,897 4.2 49,883 2.9 83,918 3.4 89,798 4.2 49,883 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0							0 25642 1 1 0 7935 2 2 0 12641 1 0 48378 0 0 8678 2 0 30116 2 2 0 3548 2 0 41786 0 2668 0 1851 2 0 1360 1	5662 15.2 7935 21.4 2641 11.4 8876 13 9678 21.4 10116 23.7 3548 26.3 1786 13 2688 10 3357 9.3 1861 26.4 2740 11.9 1360 10.9	15.2 63.6 21.4 24.1 11.4 38.7 13 58 21.4 24.1 23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	4.9 99.897 4.2 49,883 2.9 83,918 3.4 89,798 4.2 49,883 3.4 89,798 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 0 0 0 0 0 0 1 0	0 0 0 0 0 0 1	0 0 0 0 1 1 0 1 1 1 1 1 1 0			0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 7935 2 2 12641 1 0 48378 0 8678 2 0 30116 2 0 3548 2 0 41786 0 2688 0 3357 0 1861 2 0 2940 1 1 3363 1	7935 21.4 2641 11.4 8878 13 8678 21.4 9016 23.7 3548 26.3 1786 13 2686 10 3357 9.3 18641 26.4 2940 11.9 1360 10.9	21.4 24.1 11.4 38.7 13 58 21.4 24.1 23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	4.2 49,883 2.9 83,918 3.4 89,798 4.2 49,883 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1	0 0 0 0 0 0 0 1 0	0 0 0 0 0 0 1	0 0 0 0 1 1 0 1 1 1 1 1 1 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0			0 12641 1 0 48379 0 8678 2 0 30116 2 0 3548 2 0 41786 0 2688 0 3357 0 1861 2 0 2940 1 0 1360 1	2641 11.4 8378 13 8678 21.4 0016 23.7 3548 26.3 1736 13 2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	11.4 38.7 13 58 21.4 24.1 23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	2.9 83,918 3.4 89,798 4.2 49,883 3.4 49,471 4.5 44,677
0 0 0 1 1 1 1 1 1 1 1 1 0 0 1 1 1 1 0 0 1 1 1 0 0 1	0 0 0 0 1 0	0 0 0 0 1	0 0 0 1 1 0 0 1 1 1 1 1 1 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 48378 0 86778 2 0 30116 2 0 417786 0 2688 0 3587 0 1861 2 0 1861 2 0 2940 1 0 1360 1	13 13 8678 21.4 0016 23.7 3548 26.3 1786 13 2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	13 58 21.4 24.1 23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	3.4 89,798 4.2 49,883 3.4 49,471 4.5 44,677
	0 0 0 1	0 0 0 0 1	0 0 0 1 0 0 1 1 1 1 1 0		0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0			0 48378 0 8678 2 0 30116 2 0 3588 2 0 41786 0 2688 0 3387 0 1861 2 0 2640 1 0 1360 1	13 13 8678 21.4 0016 23.7 3548 26.3 1786 13 2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	21.4 24.1 23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	4.2 49,883 3.4 49,471 4.5 44,677
0 0 1 1 1 1 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1 1 1 0 0 1 1 1 1 1 1 1 1 1 1 1 0 0 1	0000	0 0 0 1 0	0 0 1 1 1 1 1 1 1 0		0 0 0 0 0 0 1 0 0 0 0 0 0 0 0 1 0 0 0 0			0 8678 22 0 30116 2 0 41786 2 0 41786 0 2688 0 3357 1861 2 0 2940 1 0 1360 1	8678 21.4 0016 23.7 3548 26.3 1786 13 2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	21.4 24.1 23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	4.2 49,883 3.4 49,471 4.5 44,677
	0 0 1 0	0 0 1 0	0 0 1 0 0 1 1 1 1 1 1 0	0 0 1 0 0 0 1 0 1 0 1 0 1 0 1 0	0 0 0 0 0 0 0 1 0 0 0 0 0 0 1 0 0 0 0			0 30116 22 0 3548 22 0 41786 0 2688 0 3357 0 1861 2 0 2940 1 0 1360 1	0116 23.7 3548 26.3 1786 13 2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	23.7 30.8 26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	4.5 44,677
	0	0 1 0	0 1 0 1 1 1 1 1 1 0		0 0 0 1 0 0 0 0 0 1 0 0 0 0			0 3548 2 0 41786 0 2688 0 3357 0 1861 2 0 2940 1 0 1360 1	3548 26.3 11786 13 2668 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	26.3 21.3 13 58 10 33.7 9.3 24.8 26.4 19.9	4.5 44,677
	0 1 0 1 1 1 1 1		1 1 1 1 1 0		0 0 0 1 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 41786 0 2688 0 3357 0 1861 2 0 2940 1 0 1360 1	11786 13 2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	13 58 10 33.7 9.3 24.8 26.4 19.9	90 700
	1 0 0 1 1 1 1 1	1 0 0 1 1 1 1	1 1 1 1 1 0		0 0 0 1 0 0 0 0	0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 2688 0 3357 0 1861 2 0 2940 1 0 1360 1	2688 10 3357 9.3 1861 26.4 2940 11.9 1360 10.9	10 33.7 9.3 24.8 26.4 19.9	3.4 89,798
	0 0 1 1 1 1	0 0 1 1 1	1 1 1 1 1 0		0 0 0 1 0 0 0 0	0 0	0 0 0 0 0 0 0 0 0 0 0 0	0 3357 0 1861 2 0 2940 1 0 1360 1	3357 9.3 1861 26.4 2940 11.9 1360 10.9	9.3 24.8 26.4 19.9	4.2 81,152
	0 1 1 1 1 1	0	1 1 1 1 1 0	0 0 1 0 1 0 1 0 1 0 1 0 1 0 0 0 0	0 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0	0 1861 2 0 2940 1 0 1360 1	1861 26.4 2940 11.9 1360 10.9	26.4 19.9	2.6 76,554
	0 1 1 1 1 1	0	1 1 1 1 1 0	1 0 1 0 1 0 1 0 1 0 1 0 0 0	1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 0 0		0 0 0 0 0 0 0	0 2940 1 0 1360 1	2940 11.9 1360 10.9		
	1 1 1 1	1	1 1 1 1 0 0	1 0 1 0 1 0 0 0	1 0 0 1 0 0 1 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0	0 0 0	0 1360 1	1360 10.9		5.5
	1 1 1	1	1 1 1 0 0	1 0 1 0 1 0 0 0	1 0 0 1 0 0 1 0 0 0 0 0 0	0 0 0	0 0				
	1	1	1 1 0 0	0 0	0 0 0	0 0 0					3.7
	1	1	1 1 0 0	0 0	0 0 0		0 0				4.2 69,079
	1		1 0 0	0 0	0 0 0			0 2446	2446 9.6		3.5 92,118
		- 1	0				0 0	0 1651	1651 9.6	9.6 42.3	3.5 92,118
1 1 0 0 1 1 1 0 1 1 0 1 1 1 1 1 1 1 1 1	0	0	0	0 0	0 0 0	0 0 0	0 0	0 1205 1		11.7 19.7	4 63,777
	0	0				0 0 0				11.7 19.7	4 63,777
	1	1	1	1 0	1 0 0		0 0	v 1205 1	1205 11.7		2.9 54,735
		1	1	1 0		00				9.8 41.4	4.2 97,076
					1 0 0		0 0	0 2520	2520 9.8	0.0	
			0	0	1 0 0	0 0 0	0 0	0 2520 0 5685	2520 9.8 5685 7.2	7.2 36.5	
				0 0	0 0 0	0 0 0	0 0 0 0 0 0	0 2520 0 5685 0 1267 1	2520 9.8 5685 7.2 1267 10.5	7.2 36.5 10.5 39.6	3.2 76,923
1 0 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0			0	0 0	0 0 0 0 0 0	0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0	0 2520 0 5685 0 1267 1 0 2208	2520 9.8 5685 7.2 1267 10.5 2208 9.4	7.2 36.5 10.5 39.6 9.4 24.6	3.2 76,923 3.4 71,611
1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0			0	0 0 0 0	0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0	0 2520 0 5685 0 1267 1 0 2208	2520 9.8 5685 7.2 1267 10.5 2208 9.4 2208 9.4	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6	3.2 76,923 3.4 71,611 3.4 71,611
0 1 0 1 1 1 1 1 1 1 1 1 0 0			0 0 0	0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5685 0 1267 1 0 2208 0 2208 0 1841 1	2520 9.8 5685 7.2 1267 10.5 2208 9.4 2208 9.4 1841 18.3	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6 18.3 24.3	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749
0 1 0 1 1 1 1 1 1 1 1 1 0 0	0		0 0 0	0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 2520 0 5685 0 1267 1 0 2208 0 2208 0 1841 1 0 11040 1	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1281 18.3 0404 11.1	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6 18.3 24.3 11.1 35.9	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668
1 0 1 1 1 1 1 1 1 1 1 0 0	0	A	0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0		0 2550 0 5665 0 1267 1 0 2208 0 2208 0 1841 1 0 110404 1 0 25284	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1241 18.3 0404 11.1 6284 9	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6 18.3 24.3 11.1 35.9 9 37.5	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 3 85,968
0 1 1 1 1 1 1 1 1 1 0 0	1	0	0 0 0 0			0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5685 0 1267 1 0 2208 1 0 2208 1 1 1040 1 1 10404 1 0 25284 0 2243 1	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 46284 9 2243 11.5	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6 18.3 24.3 11.1 35.9 9 37.5 11.5 39.5	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,666 3 85,666 2.9 77,369
1 0 1 1 1 1 1 1 1 0 0	0	1	0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5685 0 1267 1 0 2208 1 0 2208 1 1 1040 1 1 10404 1 0 25284 0 2243 1	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 6624 9 2243 11.5	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6 18.3 24.3 11.1 35.9 9 37.5 11.5 39.5 7.7 38.1	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 3 85,966 2.9 77,369 1.9 84,899
1 0 1 1 1 1 1 1 1 0 0	0	1	0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 1 0	0 1 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5665 0 1267 1 0 2208 0 1941 1 0 110040 1 0 26284 0 2243 1 0 2135	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 46284 9 2243 11.5 2135 7.7	7.2 36.5 10.5 39.6 9.4 24.6 9.3 24.6 18.3 24.3 11.1 35.9 9 37.5 11.5 39.5 7.7 38.1 5.3 56.9	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 3 85,968 2.9 77,369 1.9 84,898 2.4 99,174
1 0 1 1 1 1 1 1 1 0 0	0	1	0 0 0 0 0 0 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0	0 1 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 5665 0 1267 1 0 2208 0 12641 1 0 110404 1 0 2268 0 2284 0 2283 1 0 2233 1 0 2135	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 45284 9 22135 7.7 6988 5.3	7.2 36.5 10.5 39.6 9.4 24.6 9.4 24.6 9.4 24.6 9.4 24.6 9.4 24.6 9.37.5 11.1 11.5 39.5 7.7 38.1 5.3 56.9 9 17.5	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,666 2.9 77,369 2.9 77,369 1.9 84,896 2.4 93,74 3.8 88,757
1 1 1 1 1 1 0 0	1	1 0 0	0 0 0 0 0 0 1 0 0	0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0	0 1 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2550 5685 0 1267 1 0 2208 0 2208 0 1841 1 0 110404 1 0 2263 0 2231 0 2235 0 6998 0 4215	9.8 9.8 5685 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 5224 9 2243 11.5 2135 7.7 6988 5.3 4215 9	7.2 36.5 10.5 33.6 9.4 2.4.6 18.3 2.4.3 11.1 35.9 9 37.5 11.5 39.5 7.7 38.1 5.3 56.9 9 17.5 14.8 38.1	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 9,174 3.8 86,757 3.9 63,822
1 1 1 1 1 1 0 0	0	1 0 0	0 0 0 0 0 0 1 0 0		0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5665 0 1267 1 0 2208 0 1841 1 0 110404 1 0 2243 1 0 2243 0 2243 0 2243 0 2245 0 4245 0 4225 0 1170 1	2520 9.8 5685 7.2 1267 10.5 2208 9.4 2208 9.4 1841 18.3 0404 11.1 6284 9 2243 11.5 2135 7.7 6988 5.3 91170 14.8	7.2 36.5 10.5 39.6 9.4 24.6 11.1 35.9 9 37.5 11.5 39.5 7.7 38.1 9 37.5 14.8 38.1 14.8 38.1	3.2 76,923 3.4 71,611 3.4 71,749 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 4.2 66,071
1 1 1 1 1 0 0		1 0 0 0	0 0 0 0 1 0 0 0 0 0 0 1		0 0 0 0 0 0			0 2550 0 5685 0 1267 1 0 2208 0 2208 0 1841 1 0 11044 1 0 2263 0 2231 0 2253 0 4215 0 4215 0 1354 1	2520 9.8 5605 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 5624 9 2203 11.5 2135 7.7 5688 5.3 4215 9 1170 14.8 1354 13.6	7.2 36.5 10.5 30.6 9.4 24.6 18.3 24.3 11.1 35.9 9 37.5 11.5 39.5 15.3 56.9 9 17.5 14.8 38.1 13.6 21.2	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 9,174 3.8 88,757 3.9 63,822 4.2 66,071 2.9 69,556
1 1 1 1 1 0 0		1 0 0	0 0 0 0 1 0 0 0 0 0 0 1 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5685 0 1267 1 0 2208 0 3264 0 110404 1 0 2263 1 0 2263 0 2243 1 0 2233 1 0 6988 0 4225 0 1170 1 0 1354 1 0 233 3	2520 9.8 5685 7.2 1267 10.5 2208 9.4 1841 18.3 0404 11.1 6284 9 22135 7.7 6988 5.3 4215 9 1170 14.8 1354 13.6 2313 10.3	7 2 365 10.5 386 94 246 9.4 246 246 246 11.1 359 357 351 11.5 39.5 361 361 7.7 38.1 38.1 36.3 53.3 56.9 9 17.5 14.8 38.1 36.1 36.2 21.2 10.3 22.2 10.3 22.2 10.3 22.2 10.3 22.5 1 10.4 25.1 1 1 1 1 36.3 1 2.5 1 <	3.2 76,923 3.4 71,611 3.4 71,749 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 4.2 66,071 2.9 9,566 3.8 57,702
0	0	1 0 0 0 1 0 0	0 0 0 0 0 1 1 0 0 0 0 1 0 0 0		0 0			0 2550 0 5685 0 1267 0 2208 0 2208 0 13441 0 2263 0 2243 0 2235 0 4215 0 1354 0 1354 0 2333 0 2233	9.8 9.8 5665 7.2 1267 10.5 2206 9.4 1841 18.3 0404 11.1 5238 9.2 2243 11.5 2235 7.7 6988 5.3 4215 9 1170 14.8 1354 13.6 2313 10.3 22277 12.6	7 2 365 105 336 9.4 246 9.4 246 111 359 9 375 115 395 115 395 115 395 9 375 115 395 9 175 148 38.1 153 569 9 175 148 38.1 136 212 103 282 126 25.1	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 3 65,968 2.9 77,369 1.9 84,898 2.4 97,174 3.8 68,757 3.9 63,322 4.2 66,071 2.9 97,566 3.8 57,702 3.9 63,822
0	1	1 0 0 0 1	0 0 0 0 0 1 1 0 0 0 0 1 0 0 0			0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5685 0 1267 1 2206 0 2208 0 12624 0 2208 0 12641 0 2263 0 2283 0 6968 0 4215 0 1354 0 2333 0 2277 1 0 1754 1	2520 9.8 2520 9.4 10.5 9.4 1247 10.5 2208 9.4 1841 18.3 0404 11.1 4526 9 1215 7.7 6488 5.3 4215 9 1170 14.8 2313 10.3 2231 10.3 2231 10.3 2231 10.3 2247 12.6 2774 14.8	7 2 365 10.5 39.6 94 24.6 9.4 24.6 24.8 24.8 18.3 24.3 31.1 35.9 9 37.5 31.5 35.5 53.9 9 37.5 11.5 35.5 59.9 11.7 13.6 21.2 11.3 11.3 24.2 12.6 12.1 10.3 24.2 12.6 25.1 12.4 83.1 1	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 4.2 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 78,091
0		1 0 0 1 0 0 0 0 0	0 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 1		0 0 0 0 0 0			0 2550 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 4215 0 11374 0 2333 1 3544 0 2333 1 2777 1 2777 0 2754	2520 9.8 565 7.2 1267 10.5 2208 9.4 1841 18.3 1841 18.3 2233 11.5 2243 11.5 2233 11.5 31354 13.6 31354 13.6 32313 10.3 2277 12.6 1754 14.8 72692 8.7	7 2 365 105 336 9.4 246 9.4 246 11.1 359 9 375 11.5 335 11.5 35 11.5 35	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 91,74 3.8 88,757 3.9 63,022 4.2 66,071 2.9 9,0566 3.8 57,702 3.9 63,822 2.9 78,091 2.9 6,056 3.8 57,702 3.9 63,822 2.2 78,091 2.8 106,743
0		1 0 0 1 0 0 0 0 1 1 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 2520 0 5685 0 1267 0 2206 0 2208 0 1841 0 110464 0 2243 0 2243 0 2135 0 6988 0 1170 1 1354 1 22277 1 2333 0 22952 0 3357	2520 9.8 565 7.2 1267 10.5 1267 10.8 2208 9.4 2208 9.4 2208 9.4 2208 9.4 9 9.2 2203 11.5 2135 7.7 9 1234 110 14.8 1354 13.6 2213 10.3 2227 12.6 2233 10.3 2227 12.6 7.7 12.6 2233 10.3 2277 12.6 7174 14.8 2892 8.7 3357 6.9	7 2 36.5 10.5 33.6 9.4 24.6 9.4 24.6 24.6 10.7 11.1 35.9 9 37.5 9.3 37.5 56.9 9 9.3 7.7 38.1 11.5 39.5 7.7 38.1 13.6 21.2 11.5 39.5 10.3 28.2 12.6 25.1 12.4 83.1 14.8 38.1 14.8 38.1 8.7 35.7 6.9 52.3 3 3 3 3	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 4.2 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 78,091
0	1	1 0 0 1 0 0 0 0 1 0 0 1	0 0 0 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0			0 2550 0 5685 0 1267 0 2208 0 2208 0 1341 0 110404 0 2233 0 6988 0 2243 0 2235 0 4215 0 1354 0 2333 0 2334 0 2333 0 1754 0 2357 0 3357 0 13194	2520 9.8 5695 7.2 1247 10.5 1247 10.5 20208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.7 6988 5.3 2215 7.7 6988 5.3 21315 10.3 22313 10.3 22313 10.3 22313 10.3 22313 10.3 22327 12.6 1754 14.8 20292 8.7 3357 6.9 1194 13.9	72 365 94 246 93 246 183 243 11.1 359 9 375 115 385 9375 381 53 669 9 175 148 381 103 242 103 242 103 242 103 242 103 242 103 242 103 242 103 242 104 38.1 156 212 103 242 104 38.7 105 25.1 14.8 38.1 157 53.5 159 50.3	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 57,702 3.9 63,822 2.2 78,091 2.8 106,743 5.5 78,779
0	1	1 0 0 1 0 0 0 1 1 0 1 1	0 0 0 0 1 0 0 0 1 0 0 0 1 1 0 0 0 1 1		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0	0 0 0 0 0 0		0 2520 0 5685 0 1267 0 2206 0 2208 0 12841 0 110404 0 22283 0 2283 0 6968 0 11104 0 1354 0 22777 1 0 2872 0 0 2357 0 1194 0 2738	2520 9.8 565 7.2 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 11.5 2135 7.7 2135 7.7 2131 10.3 22273 12.6 3233 10.3 2274 13.4 2802 8.7 3357 6.9 1124 13.9 1237 15.7	7 2 365 105 336 94 246 94 246 246 163 243 18.3 24.3 245 163 243 9 37.5 569 9 17.5 38.5 7.7 38.1 13.6 212 12.6 25.1 10.3 282 12.6 25.1 13.9 13.4 8.7 35.7 35.9 35.1 13.9 30.3 13.9 30.3 13.9 30.3 15.7 27.8 13.9 30.3 13.9 30.3 13.7 13.9 30.3 13.7 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 30.3 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 2.4 85,768 2.9 77,369 2.4 88,757 3.8 68,757 3.9 63,382 2.9 67,556 3.8 57,702 3.9 63,382 2.2 78,091 2.8 106,743 5.5 78,779 3.7 63,894
	1	1 0 0 1 0 0 0 1 0 1 1 1 1	0 0 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 1 0 0 0 1 1 1 1		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 1			0 2550 0 5685 0 1267 0 2208 0 2208 0 13441 0 2263 0 2243 0 2235 0 4215 0 1354 0 2333 0 2334 0 2734 0 3757 0 1374 0 3778 0 3778	2520 9.8 5655 7.2 1267 10.5 1267 10.5 20208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.1 0.3 5.3 2215 7.7 6968 5.3 2135 7.7 6978 5.3 11354 13.6 22313 10.3 223277 12.6 1754 14.8 20202 8.7 3357 6.9 1341 13.9 2728 15.7 374 11.9	7 2 365 94 246 246 93 246 246 18.3 243 245 11.1 355 355 11.5 38.5 355 11.5 36.9 17.7 38.1 38.1 36.1 9 17.5 14.8 38.1 10.3 24.2 10.3 24.2 10.3 24.2 12 10.3 24.2 10.3 24.2 5.1 14.8 38.1 12.6 25.1 14.8 36.7 6.9 52.3 15.7 27.8 30.3 15.7 27.8 11.9 45.5 5.3 5.5 5.5	3.2 76,923 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 2.9 69,566 3.8 57,702 3.9 63,822 2.2 78,001 2.8 106,743 5.5 78,779 3.7 63,894 2.7 75,041
0	1 1 1 0	1 0 0 1 0 0 1 0 1 0 1 1 0	0 0 0 0 1 1 0 0 0 1 1 0 0 0 1 1 1 1 1 1		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 1 0 0 0 0 0	0 0 0 0 0 0		0 2520 0 5685 0 1267 0 2206 0 2208 0 1841 0 24264 0 2243 0 2426 0 2243 0 2426 0 4215 0 1354 0 1354 0 2277 0 23857 0 1154 0 2738 0 3774 0 3724 0 1304	2520 9.8 565 7.2 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.5 2135 7.7 2135 9.3 2213 10.3 22277 12.6 2233 10.3 2227 12.4 2337 6.9 1134 13.9 1234 13.7 2738 15.7 3724 11.9 1304 12.1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 77,641 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 2.4 92,77,369 2.4 93,74 3.8 68,757 3.9 63,822 2.9 69,566 3.8 57,702 3.9 63,822 2.2 78,091 2.8 106,743 5.5 78,779 3.7 63,894 2.7 75,041 3.1 64,214
	1 1 1 0 0	1 0 0 1 0 0 0 1 1 1 1 1 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 0 0 0 0			0 2550 0 5685 0 1267 0 2208 0 2208 0 1341 0 110404 0 2233 0 6988 0 2235 0 4215 0 1354 0 2333 0 2334 0 2357 0 3357 0 1354 0 2738 0 3357 0 1304 0 3304	2520 9.8 5655 7.2 1267 10.5 1267 10.5 1267 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.7 698 5.3 2215 7.7 698 5.3 1354 13.6 2333 10.3 2277 12.6 1754 14.8 1354 13.6 2333 10.3 2728 15.7 7724 11.9 1304 12.1	72 365 94 246 93 246 11 353 94 246 133 243 11.1 353 93 355 17.7 38.1 5.3 669 9 17.5 14.8 38.1 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 11.6 12.1 11.9 45.5 12.1 18.1	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,671 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 78,091 2.8 106,743 5.5 78,779 3.7 63,874 2.7 75,041 3.1 64,224
0	1 1 0 0 0 0	1 0 0 1 0 0 0 1 0 1 1 1 1 0 0 0 0 0 0	0 0 0 0 1 0 0 1 0 0 1 1 0 0 1 1 1 1 1 1			0 0 0 0 0 0		0 2520 0 5685 0 1267 0 2206 0 2208 0 1841 0 110404 0 2235 0 2243 0 2243 0 2243 0 6688 0 4215 0 1134 0 2383 0 2277 1 0 23857 1134 0 2728 1 3774 1 3364 0 1304 1 3304	2520 9.8 565 7.2 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 7.7 2135 7.7 2135 7.7 2110 14.8 1334 13.3 1331 10.3 2277 12.6 337 6.9 1134 13.9 1334 13.3 1304 12.1 1304 12.1 1304 12.1 1304 12.1 1304 12.1 1304 12.1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 77,651 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 2.9 77,369 1.9 64,876 2.4 92,74 3.8 88,757 3.9 63,822 2.9 66,563 2.9 66,563 2.9 66,563 3.8 57,702 3.9 63,822 2.2 78,091 2.8 106,743 5.5 78,779 3.7 63,894 2.7 75,041 3.1 64,224 3.3 61,796
0	1 1 0 0 0 0 0	1 0 0 1 0 0 0 1 1 1 1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 1 1 1 1		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0			0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 6988 0 4215 0 1354 0 2333 0 2333 0 2738 0 1354 0 2738 0 3357 0 1304 0 3736 0 1304 0 2555 0 5596	2520 9.8 5655 7.2 1267 10.5 1267 10.5 1267 10.5 20208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.6 2233 11.5 2233 10.3 2333 10.3 2333 10.3 2277 12.6 1754 14.8 1354 13.6 2333 10.3 2728 15.7 1304 12.1 1304 12.1 2515 13.3 5576 15.7	72 365 94 246 93 246 11 350 94 246 133 243 11.1 350 9 75 115 395 7.7 38.1 5.3 669 9 175 14.8 38.1 10.3 242 10.3 242 10.3 242 10.3 242 10.3 242 11.6 21.2 10.3 242.2 10.3 242.2 11.6 35.7 6.9 52.3 11.9 45.5 12.1 18.1 13.3 20 15.7 18.8	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,671 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,024 3.5 78,779 3.7 63,874 2.7 75,041 3.1 64,224 3.3 61,708 3.3 61,708
0	1 1 0 0 0 0	1 0 0 1 0 0 0 1 1 1 1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 1 1 0 0 0 0 0 0 0 1 1 1 1 1		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0			0 2520 0 5685 0 1267 0 2206 0 2208 0 1841 0 110404 1 22534 0 22433 0 2686 0 2243 0 6968 0 11010 1 1354 0 1574 0 2387 0 11344 0 33774 1 13044 0 2555 0 1303	2520 9.8 5565 7.2 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2213 7.7 2125 7.7 2135 7.7 2110 14.8 1334 13.6 13231 10.3 2277 12.6 337 6.9 1124 14.8 2892 8.7 3724 11.9 1334 12.1 1304 12.1 1304 12.1 133 15.7 5576 15.7 1533 15.7	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 77,641 3.4 71,611 3.4 71,611 2.4 51,749 3.4 83,668 2.9 77,369 1.9 64,876 2.4 97,7369 2.4 97,369 2.4 97,369 3.0 63,822 2.4 66,071 2.9 67,702 3.9 63,822 2.9 57,702 3.9 63,822 2.7 75,041 3.1 64,224 3.1 64,224 3.3 61,708 3.5 56,866 3.7 52,220
1	1 1 0 0 0 0 0	1 0 0 1 0 0 0 1 0 1 1 1 1 0 0 0 0 0 0 0	0 0 0 0 1 0 0 1 1 0 0 1 1 0 0 1 1 1 1 1			0 0 0 0 0 0		0 2520 0 5685 0 1267 0 2206 0 2208 0 1841 0 110404 1 22534 0 22433 0 2686 0 2243 0 6968 0 11010 1 1354 0 1574 0 2387 0 11344 0 33774 1 13044 0 2555 0 1303	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2213 7.7 2135 7.7 2135 7.7 2135 7.7 2135 7.7 1235 13.6 1334 12.3 2277 12.6 2333 10.3 2277 12.6 3357 6.9 1134 13.7 3357 6.9 1134 12.1 1334 12.1 1334 12.1 1335 15.7 3103 20.1	7 2 365 105 366 94 246 9.4 246 246 246 111 35.9 35.1 11.1 35.9 9 37.5 35.1 35.5 36.9 9 17.5 14.8 38.1 36.1 36.2 21.2 10.3 28.2 11.5 35.7 6.9 9 17.5 14.8 38.1 35.7 6.9 52.3 11.4 35.7 6.9 52.3 11.9 45.5 12.1 18.1 13.3 20.3 15.7 27.8 12.1 18.1 13.3 20.1 14.0 24.1 13.4 <td< td=""><td>3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 81,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,054 3.5 78,779 3.7 63,874 3.1 64,224 3.3 61,798 3.5 56,866 3.7 52,278 3.7 52,278</td></td<>	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 81,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,054 3.5 78,779 3.7 63,874 3.1 64,224 3.3 61,798 3.5 56,866 3.7 52,278 3.7 52,278
1	1 1 0 0 0 0 0 0 0	1 0 0 1 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0			0 2530 0 5685 0 1267 0 2208 0 2208 0 1341 0 110404 0 2233 0 6988 0 2243 0 2235 0 4215 0 1354 0 2333 0 2738 0 3357 0 1354 0 2738 0 3367 0 1304 0 2738 1 3044 0 2555 1 5696 1 1903 0 1803	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.1 0.3 22.1 2215 7.7 698 5.3 22313 10.3 2243 10.3 2277 12.6 1754 14.8 3337 6.9 1304 12.1 1304 12.1 2515 13.3 557 15.7 1803 20.1	72 365 94 246 94 246 183 243 183 243 11.1 359 9 375 7.7 38.1 11.5 395 7.7 38.1 136 212 148 38.1 136 212 126 25.1 139 30.3 148 38.1 136 212 126 25.1 139 30.3 157 27.8 159 20.1 151 16.1 152 118.1 153 20 154 40.2 201 40.2 201 40.2	3.2 76,923 3.4 77,641 3.4 71,611 3.4 71,611 3.4 81,749 3.4 83,668 2.9 77,369 1.9 84,898 2.4 97,7369 1.9 84,898 2.4 97,369 3.8 88,757 3.9 63,822 2.9 66,565 3.8 57,702 3.9 63,822 2.2 78,091 2.8 106,743 5.5 78,779 3.7 63,894 2.7 75,041 3.1 64,224 3.3 61,708 3.5 56,866 3.7 52,278 3.7 52,278 3.7 52,278 3.7 52,278 3.2 67,813
1	1 1 0 0 0 0 0 0 0 0 0	1 0 0 1 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0			0 0 0 0 0 0		0 2530 0 5685 0 1267 0 2208 0 2208 0 12841 0 110404 0 2233 0 2243 0 2243 0 2243 0 4295 0 1170 0 1354 0 2313 0 2337 0 1354 0 23774 0 1304 0 2555 1 1304 0 5558 1 1803 0 1803 0 1803 0 4460	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 11.1 53 5.3 3215 7.7 7047 1.48 3354 13.6 3331 10.3 2213 12.6 2277 12.6 2277 12.6 3237 6.9 1134 12.1 1304 12.1 1304 12.1 1304 12.1 1803 20.1 1803 20.1 1803 20.1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 81,668 2.9 77,369 1.9 84,898 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,671 2.9 66,671 2.9 66,671 2.9 63,862 3.8 87,702 3.9 63,822 2.9 76,923 3.9 63,822 3.8 87,702 3.9 63,822 2.9 76,903 3.1 64,214 3.1 64,224 3.3 61,798 3.7 52,278 3.7 52,278 3.2 67,813 3.8 86,602
1	1 1 0 0 0 0 0 0 0 0 0	1 0 0 1 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0					0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 6988 0 2233 0 4225 0 1354 1 2333 0 2738 0 1754 0 3357 0 1304 0 3778 1 3304 1 5596 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 1 1803 <tr td="" tttttt<=""></tr>	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.6 2233 10.3 2277 12.6 1754 14.8 3357 6.9 1304 12.1 1304 12.1 2515 13.3 557 13.3 1304 12.1 2515 15.7 1803 20.1 4460 12.9	72 365 94 246 94 246 183 243 111 359 9375 311 15 355 77 381 15 369 942 246 15 355 77 381 163 242 103 242 103 282 103 282 119 455 121 181 133 203 157 188 201 402 201 402 142 339	3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 81,769 3.4 83,668 2.9 77,369 1.9 84,876 3.0 63,822 3.0 63,822 3.0 63,822 3.0 63,822 3.0 63,822 2.2 78,091 2.8 106,743 5.5 78,799 3.7 63,262 3.3 61,774 3.4 64,214 3.5 56,866 3.7 52,278 3.7 52,278 3.7 52,278 3.7 52,278 3.8 68,962 2.9 78,849
0	1 1 0 0 0 0 0 0 0 0 0	1 0 0 1 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0			0 0 0 0 0 0		0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 2253 0 2233 0 2254 0 2253 0 4245 0 1354 0 2313 0 2333 0 2373 0 1354 0 2374 0 3377 0 1364 0 2555 1 1603 0 1803 0 1803 0 4460 1 15559	2520 9.8 5565 7.2 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2209 9.4 2201 11.1 330 6.3 3215 7.7 6468 5.3 3215 9 9 9 2233 10.3 105 13.5 125 12.6 1274 14.8 2802 8.7 3724 15.7 3724 15.7 3724 15.7 333 6.9 1134 12.1 1334 12.1 1303 20.1 1803 20.1 1531 14.2 9599 7.7	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	$\begin{array}{cccccccccccccccccccccccccccccccccccc$
	1 1 1 0 0 0 0 0 0 0 0 1 1 1	1 0 0 0 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 6988 0 4215 0 1354 1 2333 0 2333 0 2738 0 1354 0 2738 0 3357 0 1304 0 3255 1 3344 0 2255 1 1304 0 5986 1 1803 1 1803 1 1803 1 1831 0 4460 1 1559 1 1531 0 1221	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 698 5.3 2215 7.7 698 13.4 1334 13.6 2333 10.3 2277 12.6 1754 11.9 1304 12.1 2515 13.3 557 13.3 1304 12.1 2515 15.7 1803 20.1 4460 12.9 959 7.7 9511 14.2 9529 7.7 120 13.7 </td <td>72 385 94 246 94 246 183 243 11.1 355 7.7 38.1 11.5 355 7.7 38.1 136 212 14.8 38.1 136 212 126 25.1 139 30.3 148 38.1 136 212 126 25.1 139 30.3 148 38.1 157 159 159 20.2 121 16.1 121 16.1 121 16.1 121 16.1 121 16.2 121 16.2 121 16.2 121 16.2 122 30.3 124 20.2 125 34.2 126 34.2 127 36.5</td> <td>3.2 76,923 3.4 77,651 3.4 71,611 3.4 71,611 3.4 81,769 3.4 81,769 3.4 83,668 2.9 77,369 1.9 84,896 2.9 77,369 3.0 63,822 3.0 63,822 3.0 63,822 2.4 66,071 2.9 97,702 3.9 63,822 2.2 78,091 2.8 106,733 5.5 78,799 3.7 63,894 2.7 75,041 3.1 64,214 3.3 61,798 3.5 56,866 3.7 52,278 3.7 52,278 3.8 68,062 2.9 118,494 5 82,455 2.2 67,813 3.8 68,062 2.9 118,494 <</td>	72 385 94 246 94 246 183 243 11.1 355 7.7 38.1 11.5 355 7.7 38.1 136 212 14.8 38.1 136 212 126 25.1 139 30.3 148 38.1 136 212 126 25.1 139 30.3 148 38.1 157 159 159 20.2 121 16.1 121 16.1 121 16.1 121 16.1 121 16.2 121 16.2 121 16.2 121 16.2 122 30.3 124 20.2 125 34.2 126 34.2 127 36.5	3.2 76,923 3.4 77,651 3.4 71,611 3.4 71,611 3.4 81,769 3.4 81,769 3.4 83,668 2.9 77,369 1.9 84,896 2.9 77,369 3.0 63,822 3.0 63,822 3.0 63,822 2.4 66,071 2.9 97,702 3.9 63,822 2.2 78,091 2.8 106,733 5.5 78,799 3.7 63,894 2.7 75,041 3.1 64,214 3.3 61,798 3.5 56,866 3.7 52,278 3.7 52,278 3.8 68,062 2.9 118,494 5 82,455 2.2 67,813 3.8 68,062 2.9 118,494 <
	1 1 1 0 0 0 0 0 0 0 0 1 1 1 1 0	1 0 0 0 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 2253 0 2233 0 2254 0 2253 0 4245 0 1354 0 2333 0 2372 0 1354 0 2372 0 1194 0 2372 0 1364 0 2555 1 100 1304 1 0 2555 1 1803 0 1803 0 1803 0 4460 1 44569 0 1250 1210 1	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.1 3201 11.3 2203 11.5 2155 7.7 6468 5.3 3154 1.6 2233 10.3 2207 12.6 2277 12.6 2233 10.3 3267 8.9 1194 11.8 1193 12.1 1304 12.1 1303 20.1 1583 20.1 1583 20.1 1583 20.1 139 14.2 159 13.7 159 13.7	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,856 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.2 78,001 2.8 106,743 5.5 78,779 3.7 63,824 2.7 75,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.2 67,813 3.8 86,662 2.9 18,494 5 82,455 2.1 64,212 3.2 83,856
	1 1 0 0 0 0 0 0 0 0 0 0 1 1 1 1 0 0	1 0 0 0 1 0 0 0 1 1 0 0 0 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 2134 0 2263 0 2233 0 4245 0 1354 0 2333 0 2738 0 2734 0 3357 0 3357 0 3354 0 3754 0 3357 0 1304 0 3734 0 3367 0 1304 0 555 0 1803 0 1803 0 1803 0 1511 0 4460 0 1521 0 14959	2520 9.8 5655 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 698 5.3 2215 7.7 698 5.3 2223 10.3 1354 13.6 2231 10.3 1354 13.6 23337 6.9 1304 12.1 2556 15.7 1803 20.1 4460 12.9 1334 12.1 1331 14.2 9569 15.7 1803 20.1 4460 12.9 137.1 1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 83,668 2.9 77,369 1.9 64,876 2.4 97,7369 1.9 64,876 2.4 97,7369 3.8 68,757 3.9 63,822 2.4 66,071 2.9 66,566 3.8 57,702 3.9 63,362 2.2 78,071 2.8 106,743 5.5 78,779 3.7 63,262 3.7 55,666 3.7 52,278 3.7 52,278 3.7 52,278 3.8 68,662 2.9 11,8,494 5 82,455 5 82,455 3.2 83,856 3.2 83,856 3.2 83,856 <
	1 1 1 0 0 0 0 0 0 0 1 1 1 1 0 0 0 0	1 0 0 1 0 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 2233 0 2243 0 2235 0 4245 0 1354 0 1354 0 2333 0 2373 0 1354 0 2374 0 3367 0 1304 0 2555 1 100 1304 1 0 2555 1 1803 0 1581 1 1803 0 4406 1 1191 0 1959 12100 1959 12101 1 0 1959	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.1 321 5.3 2203 11.5 2155 7.7 4668 5.3 2157 12.6 2273 12.6 2203 1.3 3254 1.5 2207 12.6 2207 12.6 2207 10.5 3237 0.9 1134 12.1 1334 12.1 1334 12.1 133 13.3 5596 15.7 1331 14.2 139 14.2 199 119 119 119 199 10.8 </td <td>$\begin{array}{cccccccccccccccccccccccccccccccccccc$</td> <td>3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 81,769 3.4 83,668 2.9 77,369 2.4 97,374 3.8 88,757 3.9 63,822 4.2 66,071 2.9 77,020 3.0 63,822 2.2 76,051 3.6 67,702 3.7 63,864 2.7 75,041 3.1 64,214 3.3 61,793 3.5 56,666 3.7 52,278 3.7 52,278 3.7 52,278 3.8 66,662 2.9 18,464 5 82,455 3.2 67,813 3.2 83,866 3.2 83,866 3.2 83,866 5 62,451</td>	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 81,769 3.4 83,668 2.9 77,369 2.4 97,374 3.8 88,757 3.9 63,822 4.2 66,071 2.9 77,020 3.0 63,822 2.2 76,051 3.6 67,702 3.7 63,864 2.7 75,041 3.1 64,214 3.3 61,793 3.5 56,666 3.7 52,278 3.7 52,278 3.7 52,278 3.8 66,662 2.9 18,464 5 82,455 3.2 67,813 3.2 83,866 3.2 83,866 3.2 83,866 5 62,451
	1 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 1 0 0 0 1	1 0 0 0 1 0 0 0 1 1 1 0 0 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 6988 0 4215 0 1334 0 2333 0 2738 0 3357 0 3357 0 3357 0 1304 0 3734 0 3364 0 5568 0 1304 0 5568 0 1803 0 1803 0 1803 0 1803 0 1803 0 1191 0 1191 0 120 0 1321 0 44569 0 120 0 1591	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 698 5.3 2215 7.7 698 5.3 2223 10.3 1354 13.6 2323 10.3 2277 12.0 1754 11.9 1304 12.1 2556 15.7 1803 20.1 4460 12.9 1303 14.2 9599 7.7 1199 10.8 4842 9.1 1599 10.8	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 83,668 2.9 77,369 1.9 64,876 2.4 97,7369 1.9 64,876 2.4 97,7369 3.8 68,757 3.9 63,822 2.4 66,071 2.9 66,566 3.8 57,702 3.9 63,362 2.2 78,071 2.8 106,743 5.5 78,779 3.7 63,262 3.7 55,666 3.7 52,278 3.7 52,278 3.7 52,278 3.8 68,662 2.9 11,8,494 5 82,455 5 82,455 3.2 83,856 3.2 83,856 3.2 83,856 <
0	1 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 1 0 0 1 0	1 0 0 1 0 0 0 0 1 1 1 1 0 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 4245 0 4251 0 1354 0 2373 0 2373 0 1354 0 2373 0 2754 0 2374 0 3357 0 1364 0 3724 0 1364 0 2555 1 1603 0 1583 0 1583 1 1803 0 4460 1 1193 0 1959 12100 1989 13842 1999 1399 1999	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 6498 9 2243 11.5 1353 10.3 1054 13.6 1354 13.6 1353 10.3 2777 12.6 7778 15.7 1304 12.1 1304 12.1 1304 12.1 1304 12.1 1303 20.1 1803 20.1 1803 20.1 199 19.9 199 10.8 199 10.8 199 10.8 199 10.8 199 10.	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 66,071 2.9 66,388 3.9 63,822 3.9 63,822 3.9 63,822 3.9 63,822 2.9 76,054 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.2 67,813 3.8 8,6%02 2.9 18,404 5 82,655 2.1 64,212 3.3 6,7813 3.8 8,6%02 2.9 18,404 5 82,655
1	1 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 0 0 0 0	1 0 0 1 0 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 2263 0 2243 0 2233 0 4245 0 1354 0 2333 0 2738 0 2734 0 3357 0 3357 0 3357 0 3364 0 3734 0 3364 0 5568 0 1304 0 5568 0 1803 0 1803 0 1803 0 1511 0 48529 0 1191 0 14959 0 4642	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 698 5.3 2215 7.7 698 10.3 1354 13.6 2231 10.3 1273 15.7 1304 12.1 2515 13.3 557 13.3 1304 12.1 2513 7.7 1931 14.2 9596 15.7 1803 20.1 4460 12.9 1959 10.8 4842 9.1<	72 365 94 246 94 246 183 243 11.1 359 94 246 183 243 11.1 359 9375 365 11.5 395 7.3 860 9375 115 11.6 321 11.3 321 11.3 22 10.3 282 10.3 282 11.6 38.1 11.3 36.9 12.2 22.1 12.6 25.1 12.6 25.1 13.7 35.7 13.9 303 15.7 27.8 12.1 18.1 13.3 20 14.0 2.2 20.1 40.2 12.2 34.3 13.7 35.5 11.9 28.5 14.2 39.9 <td>3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.2 67,813 3.8 88,662 2.9 18,404 5 82,655 2.1 64,212 3.3 61,798 3.2 7,781</td>	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.2 67,813 3.8 88,662 2.9 18,404 5 82,655 2.1 64,212 3.3 61,798 3.2 7,781
1	1 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 1 0 0 1 0	1 0 0 1 0 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 2243 0 2253 0 2243 0 2254 0 2253 0 4265 0 1354 0 2333 0 2373 0 1354 0 2374 0 2382 0 3364 0 2783 0 3724 0 3364 0 3559 0 1304 0 5598 0 1803 0 1803 0 1989 1210 1 0 1959 1210 1 0 15677 0 3647 0 14492 0 14492 <	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.5 215 7.7 6468 5.3 3215 9 9 9 2233 10.3 1054 1.6 1355 10.5 2233 10.3 2267 12.6 277 12.6 278 15.7 7724 15.7 7728 15.7 7324 15.7 7324 12.1 1304 12.1 1303 20.1 1583 20.1 1597 12.0 1303 20.1 157 12.1	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 83,668 2.9 77,369 1.9 84,896 2.9 77,369 3.0 63,822 4.2 66,071 2.9 64,866 2.9 77,02 3.0 63,822 2.2 76,051 2.8 106,717 2.8 106,713 5.5 78,779 3.7 63,864 3.3 61,707 3.4 64,214 3.5 56,866 3.7 52,277 3.7 52,2778 3.7 52,2778 3.8 68,663 2.9 18,844 5 82,455 3.4 64,212 3.2 63,322 3.3 63,227 <
1	1 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 0 0 0 0							0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 11044 0 2233 0 4245 0 4245 0 4251 0 3154 0 2277 0 2754 0 2754 0 2754 0 2754 0 2754 0 2754 0 3754 0 3754 0 3754 0 3754 0 3744 0 3545 0 1304 0 1304 0 1304 0 1303 0 1803 0 1803 0 1989 0 1989 1989 1	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 648 9 2215 7.7 6498 5.3 22071 12.6 22771 12.6 22771 12.6 2323 0.9 2313 10.3 2502 8.7 2323 0.9 1154 13.9 1154 13.9 1154 13.9 1154 13.9 1154 13.7 1156 13.7 1557 13.3 5596 15.7 153 14.2 159 10.8 159 10.	72 365 94 246 183 243 111 359 94 246 183 243 111 359 91 246 183 243 935 115 935 115 94 246 935 115 935 115 94 246 94 246 91 75 148 381 362 22 103 282 126 251 148 381 37 77 39 303 157 28 20.1 402 21 18.1 13.3 29 14.2 33.3 7.7 98 20.1 402 22.1 402 24.2 343 13.7 <	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 66,071 2.9 66,071 2.9 66,38 3.9 63,822 3.9 63,822 3.9 63,822 2.9 70,021 3.9 63,822 2.9 70,021 3.9 63,822 3.7 63,824 2.7 75,041 3.1 64,214 3.3 61,798 3.7 52,278 3.2 67,813 3.8 8,662 2.9 18,494 5 82,645 2.1 64,212 3.3 61,798
1	1 1 1 0 0 0 0 0 0 0 0 1 1 1 1 1 0 0 1 1 1 1 1 1 1 1 1 1 0	1 0 0 0 0 0 0 0 1 1 1 1 1 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 6988 0 4215 0 1354 0 2333 0 2738 0 2734 0 3357 0 3357 0 3357 0 1304 0 2738 0 3367 0 1304 0 2515 0 1803 0 1803 0 1803 0 181 0 1803 0 1803 0 181 0 4842 0 15677 0 36677 0 1403 140337	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 1.1.5 2215 7.7 698 5.3 22235 7.7 698 1354 1354 13.6 22313 10.3 2277 12.6 1754 11.9 1304 12.1 2555 15.7 1803 20.1 4460 12.9 1959 10.8 4642 9.1 1957 12.8 3077 10 10377 1	7 2 365 94 246 366 94 246 163 94 246 163 163 243 163 11.1 359 97.7 38.1 11.5 395 7.7 38.1 11.6 11.5 395 7.7 38.1 11.6 24.2 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.6 24.2 24.8 11.7 25.5 24.8 11.9 25.5 25.5 12.9 34.3 14.8 11.7 25.5 12.9 11.7 25.5 12.9 11.7	3.2 76,923 3.4 71,611 3.4 71,611 3.4 71,611 3.4 71,611 3.4 81,769 3.4 83,668 2.9 77,369 2.4 97,369 2.4 99,374 3.8 88,757 3.9 63,322 4.2 66,071 2.9 76,623 5.5 78,779 3.7 63,864 2.7 75,041 3.1 64,214 3.1 64,214 3.3 61,796 3.4 66,602 2.9 73,207 3.7 52,278 3.7 52,278 3.7 52,278 3.8 66,603 2.9 18,464 5 82,455 2.1 64,212 3.2 67,813 3.6 63,227 3.6 63,227
	1 1 1 0 0 0 0 0 0 0 1 1 1 1 1 0 0 0 1 1 1 1 0 0 0 1 1 1 1 1 0 0 0 0 0 1 1 1 1 0							0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 2233 0 2243 0 2243 0 2243 0 2455 0 4245 0 1354 0 2333 0 2797 0 1354 0 2373 0 1754 0 2387 0 1164 0 2372 1 1304 0 2555 1 1603 0 1581 0 1583 1 1304 0 1583 1 1499 1 1507 1 1492 1 1493 0 1492 1 1493	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.1 33 5.3 10.3 11.1 2203 10.3 2204 1.4 1351 1.6 1353 10.3 2207 12.6 2203 10.7 2203 1.7 1204 13.8 3270 15.7 7724 15.7 7728 15.7 7728 15.7 73724 11.9 1205 13.3 5596 15.7 1303 20.1 1303 20.1 1303 20.1 1527 <t< td=""><td>72 365 94 246 18.3 243 11.1 359 94 246 18.3 243 11.1 359 9 75 11.5 38.5 77 38.1 5.3 66.9 9 17.5 14.8 38.1 10.3 242 10.3 242 10.3 242 11.6 7.7 12.1 13.6 12.2 22.5 12.4 38.7 6.9 52.3 13.9 30.3 15.7 12.8 20.1 402 20.1 402 21.1 18.1 13.7 35.5 11.9 23.5 11.9 28.2 12.8 20.7 18.9 25.5 11.9 25.5 11.9 24.4 <!--</td--><td>3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.8 86,602 2.9 18,404 5 82,656 2.1 64,212 3.8 88,602 2.9 18,404 5 52,278</td></td></t<>	72 365 94 246 18.3 243 11.1 359 94 246 18.3 243 11.1 359 9 75 11.5 38.5 77 38.1 5.3 66.9 9 17.5 14.8 38.1 10.3 242 10.3 242 10.3 242 11.6 7.7 12.1 13.6 12.2 22.5 12.4 38.7 6.9 52.3 13.9 30.3 15.7 12.8 20.1 402 20.1 402 21.1 18.1 13.7 35.5 11.9 23.5 11.9 28.2 12.8 20.7 18.9 25.5 11.9 25.5 11.9 24.4 </td <td>3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.8 86,602 2.9 18,404 5 82,656 2.1 64,212 3.8 88,602 2.9 18,404 5 52,278</td>	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.8 86,602 2.9 18,404 5 82,656 2.1 64,212 3.8 88,602 2.9 18,404 5 52,278
1	1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 1 1 1 1 0 0 0 0 0 1 1 1 0 0 0 1 1 0	1 0 0 1 0 0 0 0 1 1 1 1 1 1 1 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 1841 0 110404 0 2233 0 6988 0 4215 0 1354 0 2333 0 2738 0 2734 0 3357 0 3357 0 3357 0 1304 0 2738 0 3367 0 1304 0 2555 0 1803 0 1803 0 1803 0 1803 0 1803 0 1803 0 1803 0 1803 0 1803 0 1803 0 1803 0 1807	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 1267 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2203 7.7 698 5.3 2215 7.7 698 1354 1354 13.6 22313 10.3 2277 12.6 1754 11.9 1304 12.1 2555 15.7 1803 20.1 4460 12.9 1304 12.1 157 11.9 1304 12.1 1581 14.2 959 10.8 4642 9.1	72 365 94 246 94 246 183 243 183 243 11.1 359 9 375 7.7 381 11.3 359 9 175 9 175 9 175 103 222 126 251 139 30.3 148 38.1 136 212 126 25.1 139 30.3 148 38.1 157 77.8 159 30.2 159 30.3 150 40.2 201 40.2 201 40.2 201 40.2 213 18.1 117 35.5 118 20.7 119 35.5 119 32.5 119 32.5 1	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 2.4 97,369 2.9 77,369 2.4 99,374 3.8 88,757 3.9 63,822 4.2 66,071 2.9 77,020 3.0 63,822 2.9 77,02 3.0 63,822 2.2 78,071 3.3 64,322 3.4 64,714 3.5 56,566 3.7 52,270 3.7 52,270 3.7 52,270 3.2 67,813 3.8 68,962 2.9 63,429 3.6 63,221 3.6 63,221 3.6 63,265 3.7 52,505 3.8 69,804 3.6 63,605
1	1 1 1 0 0 0 0 0 0 0 0 1 1 1 1 1 0 0 0 0	1 0 0 1 1 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0						0 2530 0 5685 0 1267 0 2208 0 2208 0 2243 0 2253 0 2243 0 2254 0 4255 0 4245 0 1354 0 2333 0 2373 0 1354 0 2373 0 2774 0 3337 0 1364 0 3724 0 3603 2 1304 0 1555 1 1603 0 1581 1 1583 1 1599 1 1507 1 1482 0 1507 1 3607 1 3607 1 3403 1 3607	2520 9.8 5565 7.2 1267 10.5 1267 10.5 1267 10.5 1267 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2208 9.4 2201 11.5 215 7.7 6468 5.3 3251 10.3 2203 1.6 3351 10.3 2203 1.6 3353 10.9 1214 1.3.9 1226 1.3.3 7028 15.7 7238 15.7 7238 15.7 7324 1.4 1303 20.1 1803 20.1 1803 20.1 1507 12.8 1507 12.8 1507 12.8 1507 12.8 1507 <	72 365 94 246 105 356 94 246 18.3 243 11.1 350 11.3 355 11.5 38.5 11.7 38.1 5.3 66.9 9 17.5 14.8 38.1 10.3 242 10.3 242 10.3 242 11.6 21.2 11.6 7.7 11.6 7.7 12.6 25.1 14.8 38.1 3.9 30.3 15.7 27.8 13.9 30.3 15.7 18.8 20.1 40.2 20.1 40.2 20.1 40.2 20.1 40.2 21.2 34.3 13.7 35.5 11.9 28.8 20.1 40.2 22.9 32.2	3.2 76,923 3.4 71,611 3.4 71,611 3.4 51,749 3.4 83,668 2.9 77,369 1.9 84,896 2.4 99,174 3.8 88,757 3.9 63,822 2.9 66,071 2.9 69,566 3.8 87,702 3.9 63,822 2.9 76,023 3.9 63,822 2.9 76,024 3.8 87,702 3.9 63,822 2.9 76,041 3.1 64,214 3.3 61,798 3.7 52,278 3.7 52,278 3.7 52,278 3.8 86,602 2.9 18,404 5 82,656 2.1 64,212 3.8 88,602 2.9 18,404 5 52,278
	0 0 0 1 1 1 1 0 1 1 0 0 1 0 0 0							1 1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 0 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	1 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 0 0 1 1 0 0 0 0 1 1 0 0 0 0 0 1 1 0 0 0 0 0 1 1 1 0 0 0 0 0 0 1	1 1 0 0 2520 1 0 0 6665 1 0 0 0 2208 0 0 0 0 2208 0 0 0 0 2208 0 0 0 0 1841 1 0 0 0 2238 0 0 0 0 2233 1 0 0 0 2235 1 0 0 0 2235 1 0 0 0 2335 1 0 0 0 1354 0 0 0 0 1354 1 0 0 0 1354 1 0 0 0 2377 1 0 0 0 1354 1 1 0 0 2378 1 0 0	$\begin{array}{c c c c c c c c c c c c c c c c c c c $
0 0 0 0 1 0 0 0 0 0 1 0 0 0 0 1 1 1 1 1				0	0 0	0	0		0		0 2520 9.8 41.4
0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 1 1 1 0				0	0 0	0 0				0 1205	0 1205 11.7 19.7

	sed# ProtestorsBiden (Blue GovernorBlue	County BlueStateBlue						onUnemplo		sehold IncomePrivate Universi
0	0	54.11	0	1	1		0	0	0 12				135,528
0	0	37.52	0	0			0	0		184 9			62,910
0	0	81.21	0	0	1		0	0		660 20			
0	0	53.34	0	1	1		0			435 6			116,796
0	0	38.35	0	0			0	0			9 37.		85,968
0	0	53.48	0	1	1		0	0	0 1	580 9			106,047
0	0	29.1	0	1	1	1	0	0	0 1	151	5 24.		55,466
0	0	55.94	0	0	0		0	0			16 33.		68,748
0	0	13.19	0				0	0		315 14			53,455
0	0	92.15	0	1	1		0	0		635 15			99.897
0	0	71.47	0				0	0		091 7			
				1				-					118,494
1	1 4	64.48	0	1	1		0	0		705 16			57,660
0	0	23.81	0				0			695 8			59,491
0	0	23.81	0	0	0		0	0		695 8	.9 23.		59,491
0	0	44.23	0	1	1	1	0	0	0 4	229 12	.6 30.	5 3.1	60,761
0	0	71.41	0	0	0	0	0	0	0 1	232 10	.2 55.	5 3.3	95,151
0	0	86.42	0	1	1		0	0		219 16			95.514
0	0	36.91	0	0	0		0	0		645 11			62,120
0	0	33.11	0	-	1		0	0		438 13		0.11	64,842
0	0	74.22	0	1	1		0	0		629 13			76,614
1			0				0	0					
		50.27		1				0			11 46.		71,825
0	0	50.27	0	1			0	0		118	11 46.	5.7	71,825
0	0	57.76	0	0			0	0		992 6			117,326
0	0	53.04	0				0	0		602 9	.5 38.		91,713
0	0	43.16	0			0	0	0	0 3		3 31.	1 3.1	65,967
0	0	43.16	0				0	0			3 31.		65,967
0	0	41.47	0				0	0		614 20			52,286
0	0	30.6	0	0	-		0						53,040
1	2		0				0	0					64,299
1		63.12						ő			· ·		
	1 4	63.12	0	-			0	0		280 17			64,299
0	0	34.41	0				0	0		882 17			51,622
0	0	32.74	0				0	0		627 12			59,960
0	0	56.67	0				0	0		370 14	.3 24.		66,169
1	2	63.35	0			0	0	0		885 15	.6 34.		62,776
1	0 3	63.35	0			0	0	0		885 15			62,776
0	0	51.98	0				0	0		635 14			61,028
ő	0	41.79	0				0	0		293 10			74,264
ő	0	29.43	0				0	0			20 20.		51.053
0	0	29.43	0				0	0					
								0					
0	0	67.57	0	1	1	1	0	0		310 9		0.4	108,037
0	0	57.16	0	0	0	0	0	0		575 14			68,720
0	0	73.51	0	1	1		0	0		727 15	.7 59.		72,025
0	0	51.11	0	0	0	0	0	0	0 4	740 14		1 3.1	70,203
0	0	35.23	0	0	0	0	0	0	0 2	761 7	8 36.	6 2.4	86,374
0	0	66.36	0	0	0	0	0	0	0 3	973	16 35.	9 3.8	60,808
0	0	71.63	0	-	1	-	0	0					119.667
1			0	1			0						
		87.28		1		1	0	0		20			54,735
0	0	66.68	0	0	1	0	0	0		621 10			80,645
0	0	60.52	0	1	1	1	0	0		862 12	.7 32.		74,190
0	0	66.68	0	0	1	0	0	0		668 10	.2 48.		80,645
0	0	41.45	0	0	0	0	0	0	0 1	301 13		2 2.8	70,013
0	0	56.14	0	1	1	1	0	0	0 1	536 8		5 5.1	93,475
0	0	23.84	0	0	1	0	0	0			13 18.		58.170
0	0	58.22	0	0	1		0	0			3 40.		69,689
ő		59.82			0		0	0					
	0		0					0					62,509
0	0	59.82	0		-	-	0	0			33.		
0	0	50.71	0	0	1		0	0		635	18 4		55,876
0	0	33.36	0	0	1	0	0	0		000 11	.5 29.		57,563
0	0	67.01	0	1	1	1	0	0		950 8		6 4.7	96,483
0	0	57.52	0	1	0	1	0	0	0 3		9 35.	3 2.3	76,227
0	0	64.48	0				0	0		904 12			70,834
0	0	56.3	0				0	0		12			96,798
0	0	42.95	0				0	0			.0		72,965
			0				0	ő		14			54,167
0	0	26.74						0					58,456
0	0	48.53	0				0	0		805 11			65,949
0	0	35.07	0				0	0		104 16			
0	0	27.3	0				0	0		276 11	.4 24.		
0	0	53.84	0	0	1	0	0	0		415 17	.9 22.		57,443
0	0	39.35	0	1	1	1	0	0	0 1	851 12			62,637
0	0	31.57	0	0	1	0	0	0		710 13		5 6.2	84,738
0	0	23.23	0				0	0		043 13			54,361
0	0	42.26	0	0			0	0		912 16			59,358
0	0	46.12	0	1			0	0		10			57,666
0	0		0	0			0	0		10			72.157
		45.08											
0	0	59.43	0	0			0						
0	0	81.21	0	0			0			440 20			56,385
0	0	52.98	0	1			0	0		397 11	.3 25.		86,350
0	0	49.64	0	0			0	0		407 9	.8 33.		75,468
0	0	38.52	0	0	0	0	0	0		049 18		5 2.8	55,852
0	0	42.45	0	0	0		0	0		214 10			76,163
0	0	49.66	0	0	1		0	0	0 5	145 14			56,489
0	0	60.78	0	1			0			979 8			101,621
0	0	37.62	0	0			0	0		291 13			
0	0	41.06	0	0			0	0					65,008
0	0	48.62	0	0			0	0		057 20	.1 19.		55,065
0	0	41.17	0	0			0	0		580 8			82,434
0	0	37.95	0				0			193 16	.9 25.		49,867
0	0	42.82	0			0	0	0		290 18	.6 2	6 3.3	57,070
0	0	25.29	0				0			374 11			
0	0	71.47	0				0			420 7			
0	0	56.24	0	0			0						90,801
1								0		0/0			78.808
	0 0	44.13	0	•			0						
	0	58.88	0				0			066 13			
0	0	33.3	0				0			441 8	.7 2		
0	0	21.48	0	0	0	0	0	0	0 6	180 .	15 24.		62,666
			0				0		0 1	058 15			
0	0	23.17											
0 0 0	0		0	0	1	0	0	0	0 8			8 35	75,468
0 0 0 0 0	0	49.64		0			0			928 9	.8 33.		
0 0	0		0	0	1	0	0 0 0		0 3		.8 33. .9 24.	9 3.3	

0	ProtestorsBiden Coun				County BlueStateBlue					verty RateEdu	cationUr	nemploymentN	ledian Household Inc	omePrivate Universi
0	0	30.84		0	0		0	0	0	3630	16	22.4	4.2	60,932
0	0	79.21 53.11	0	1	1		0	0	0	4020	12.8 7.2	48.6 36.5	3.6	79,432
0	0	31.86		0			0	0	0	7352	11	26.8	4.2	97,076 65,558
0	0	56.3					0	0	0	4315 3301	8.8	50.4	2.8	96,798
0	0	86.42	0		1		0	0	0	2033	0.0	39.6	4.6	95,514
0	0	27.63		0			0	0	0	22233	16.8	18.4	4.2	50,403
0	0	28.81	0				0	0	0	1297	14.4	19.9	4.7	53,775
0	0	17.01		0	0		0	0	0	6171	18.6	17.3	3.9	41,087
0	0	57.76					0	0	0	2514	6.2	56.6	2.6	117,326
0	0	32.02	0	0	0	0	0	0	0	2476	9.4	37.5	2.5	70,837
0	0	40.23					0	0	0	8698	5.5	42.4	2.5	98,925
0	0	15.88	0	0	1	D	0	0	0	3216	22.1	18.8	5	44,153
0	0	46.12	0	1	1	1	0	0	0	1913	10.6	29.2	3.5	77,969
0	0	31.49		•			0	0	0	1074	14.6	21.5	3.4	61,105
0	0	31.15	0	0	1	0	0	0	0	3661	15.5	21.3	4	58,892
0	0	61.17					0	0	0	2125	9.7	46.9	3.7	52,278
0	0	61.17					0	0	0	2125	9.7	46.9	3	79,609
0	0	54.2	0	1	1		0	0	0	5959	13.2	22.9	4.7	78,779
0	0	54.11	0	1	1		0	0	0	16703	5.8	48.7	3.1	135,528
0	0	47.58					0	0	0	1741	9.6	32.9	2.9	73,492
0	0	57.39		0			0	0	0	1508	18.2	28.5	4.2	57,971
0	0	71.03	0	1			0	0		11878	13.7	35.5	5	82,455
0	0	74.22	0	1			0	0		20366	13.2	41.9	4.4	76,614
0	0	87.28	0	1			0	0	0	6416	20.2	35.4	2.9	54,735
0	0	33.12					0	0	0	2056	17.1	33.3	3.3	59,138
0	0	46.56					0	0	0	2337	7.9	31.5	3.2 3.5	72,032 61,729
0	0	28.57					0	0	0	1400	12.8	24.8	3.5	
0	0	55.97		0			0	0	0	3941	10.8	39.6		76,713
0	0	71.5	0	1			0	0	0	2751	11.9	45.5	2.7 2.7	75,041 59,603
0	0	48.21		0			0	0	0	2382	14.1	41.6	2.7 3.2	59,603 63,879
	0	54.83	0	1			0	0	0	1245	10.1	39.2	3.2	63,879 64,557
0	0	39.93					0	0	0	1772	12.3	25		
0	0	23.93				0	0	0	0	1762	11.8	21.6	3.1 3.1	64,025 64,025
0	0	23.93		0		0	0	0	0	1762	11.8	21.6	6.8	45,864
0	0	83.29 86.42	0				0	0	0	4863	27.7 16.5	22 39.6	4.6	45,004 95,514
0	0	86.42 67.57	0		1		0	0	0	1391	16.5 9.3	39.6 52.5	3.4	108,037
0	0	67.57 23.45		1	1	0	0	0	0	2987 2142	9.3	52.5 21.5	3.8	61,190
0	0	23.45 41.48		0			0	ő	0	1119	9.3	21.5	2.8	78,728
0	0	35.96		0		0	0	0	0	1119 1919	9.3	22.6	2.7	71,206
0	0	63.35					0	0	0	1919 4419	9.2	34.1	3.3	62,776
0	0	63.35					0	0	0	4419	15.6	34.1	3.3	62,776
0	0	28.81					0	0	0	1663	12.8	20.7	4.2	59,196
0	0	53.89	0	1	1	1	0	0	0	7797	12.0	39.7	3.3	90,740
0	0	69.07		0	1	p	0	0		14361	17.1	33.9	3.8	58,375
0	0	37.48		0		0	0	ő	0	1361	17.1	31.5	3.2	54,137
0	0	40.13					0	0		1361 13921	19.8	32.4	3.7	51,017
0	0	32.48	0	1	1	1	0	0	0	13921 1159	8.1	24.2	2.9	64,093
0	0	87.28	0	1	1	1	0	0	0	2988	20.2	35.4	2.9	54,735
ő	0	70.7	0	1	1	1	0	0	0	1164	5.6	64.5	1.7	133,068
0	0	86.42	0	1	1	1	0	0	0	2311	16.5	39.6	4.6	95,514
0	0	26.92		0)	0	0	0	0	1402	9.7	26.8	3	69,713
0	0	61.17		0)	0	0	0		12436	9.7	46.9	3.7	52,278
0	0	61.17					0	0		12436	9.7	46.9	3	79,609
0	0	53.58	0	0	1	D	0	0	0	3274	16.1	30.1	3.7	61,168
0	0	80.64	0	1	1	1	0	0	0	2441	15.1	49.9	3.2	84,548
0	0	72.44	0	1	1		0	0	0	1575	12.5	38.3	3.6	71,102
1	4	71.47	0	1	1		0	0		25286	7.7	59	2.9	118,494
0 1	0	61.2	0	1	1	1	0	0	0	1988	8.9	47.4	4.2	89,774
0	0	54.16	0	1	1	1	0	0	0	2458	8	48.5	2.1	88,032
0	0	36.1		0)	0	0	0	0	1384	8.4	31.2	1.7	79,065
0	0	36.34	0	1	1	1	0	0	0	3815	6.4	39.7	1.7	105,096
0	0	53.17		1	1	1	0	0	0	2588	13.4	30.7	4.6	67,185
0	0	26.46	0	0)	D	0	0	0	1364	13.3	29	3.3	61,798
0	0	28.63					0	0	0	1034	8.3	33.1	2	73,766
0	0	80.64	0	1	1	1	0	0	0	8938	15.1	49.9	3.2	84,548
0	0	56.46		1	1	1	0	0	0	2530	14.1	36.9	3.8	69,888
0	0	69.07	0			D	0	0	0	7616	17.1	33.9	3.8	58,375
0	0	55.51	0	0	0	D	0	0	0	6371		49	2.4	79,969
0	0	77.89			1		0	0	0	1027	11	53.1	3	143,795
0	0	61.34	0	0	0	0	0	0	0	10965	7.3	26		48,539
	0	67.57				1	0	0	0	11039	22.1	52.5	3.4	108,037
0		56.52				D	0	0	0	1053	9.3 11.5	39.5	4.2	77,369 57,971
0	0		0	0			0	0	0	1865	11.5	28.5	4.2	57,971
0 0 0	0	57.39			1	0	0	0	0	3281	18.2	29.6	3.8	97,099
0 0 0 0	0	57.39 49.66	0	0										
0 0 0 0	0 0 0	57.39 49.66 62.25	0 0	0			0	0	0	2340		56.3		
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0	57.39 49.66 62.25 63.44	0 0 0	0	1	1	0	0	0	6291	7.2	42.3	3.5	92,118
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78	0 0 0	0 1 0	1	1 D	0	0 0	0	6291 4234	7.2 9.6	42.3 39.6	3.5 2.7	92,118 81,295
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4	0 0 0 0	0	1	1 D D	0 0 0	0 0 0	0 0 0	6291 4234 2276	7.2 9.6 8.2	42.3 39.6 27.2	3.5 2.7 4.7	92,118 81,295 58,013
0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42	0 0 0 0 0	0 1 0	1	1 D D 1	0 0 0	0 0 0	0 0 0	6291 4234 2276 1091	7.2 9.6 8.2 15.3	42.3 39.6 27.2 39.6	3.5 2.7 4.7 4.6	92,118 81,295 58,013 95,514
0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5	0 0 0 0 0 0	0 1 0	1	1 D D 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0	0 0 0 0	6291 4234 2276 1091 8566	7.2 9.6 8.2	42.3 39.6 27.2 39.6 45.5	3.5 2.7 4.7	92,118 81,295 58,013
0 0 0 0 0 0 0 0 0 0 0 0 0 0 1	0 0 0 0 0 0 0 0 0 0 4	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55	0 0 0 0 0 0 0 0 0	0 1 0	1 · · · · · · · · · · · · · · · · · · ·	1 D D 1 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0	6291 4234 2276 1091 8566 21004	7.2 9.6 8.2 15.3 16.5 11.9	42.3 39.6 27.2 39.6 45.5 55.6	3.5 2.7 4.7 4.6 2.7	92,118 81,295 58,013 95,514 75,041
0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64	0 0 0 0 0 0	0 1 0	1 · · · · · · · · · · · · · · · · · · ·	1 D 1 1 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0	0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542	7.2 9.6 8.2 15.3 16.5	42.3 39.6 27.2 39.6 45.5 55.6 49.9	3.5 2.7 4.7 4.6 2.7 3.3	92,118 81,295 58,013 95,514 75,041 87,619
0 0 0 0 0 0 0 0 0 0 0 1 1 1	0 0 0 0 0 0 0 0 4 0 3	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28	0 0 0 0 0 0 0 0 0 0 0	0 1 0 1 1 1 1 1 0		1 D 1 1 1 1 0	0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.2	92,118 81,295 58,013 95,514 87,619 84,548 74,841
0 0 0 0 0 0 0 0 0 0 1 1 1 1 0 1	0 0 0 0 0 0 0 4 3 3 3	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28 65.18	0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 0 0		1 D D 1 1 1 1 1 0 D	0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325	7.2 9.6 8.2 15.3 16.5 11.9 10.5	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3	3.5 2.7 4.7 4.6 2.7 3.3 3.2	92,118 81,295 58,013 95,514 75,041 87,619 84,548
0 0 0 0 0 0 0 0 0 0 0 1 1 1 0 1 0 0 0 0	0 0 0 0 0 0 0 4 3 3 0 3 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28 65.18 65.18 41.82	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 1 0 0 0		1 D D 1 1 1 1 0 D D D	0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.2 3.9	92,118 81,295 58,013 95,514 75,041 87,619 84,548 74,841 60,571
0 0 0 0 0 0 0 0 0 0 1 1 1 0 1 1 0 0 0 0	0 0 0 0 0 0 0 0 4 0 3 3 3 0 0	57.39 40.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28 65.18 41.82 41.6	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 0 0 0 0 0 0		1 D D 1 1 1 1 0 D D D D D D D D	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2569	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.2 3.9 4.6 2.8	92,118 81,295 58,013 95,514 75,041 87,619 84,548 74,841 60,571 55,623 70,923
0 0 0 0 0 0 0 0 0 1 1 1 0 1 0 0 0 0 0 0	0 0 0 0 0 0 0 4 0 3 3 3 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28 65.18 41.82 41.6 52.74	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 0 0 0 0 0 0 0 0		1 0 1 1 1 1 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2569 2156	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 56.9	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.2 3.9 4.6	92,118 81,295 58,013 95,514 75,041 87,619 84,548 74,841 60,571 55,623
0 0 0 0 0 0 0 0 0 0 0 1 1 1 0 1 0 0 0 0	0 0 0 0 0 0 4 3 3 3 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28 65.18 41.82 41.6 52.74 61.34	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0		1 0 0 1 1 1 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 22004 1542 22122 62325 8392 2569 2156 9261	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 55.9 26	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.9 4.6 2.8 2.4	92,118 81,295 58,013 95,514 95,514 87,619 84,548 74,841 60,571 55,623 70,923 99,174
0 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 3 3 3 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.04 37.28 65.18 41.82 41.6 52.74 61.34 41.24	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2569 2156 9261 24652	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 56.9 26 34.4	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.9 4.6 2.8 2.8 2.4 4 4	92,118 81,295 58,013 95,514 75,041 87,619 84,548 74,841 60,571 55,623 70,923 99,174 48,539
0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 0 0	0 0 0 0 0 0 0 0 4 0 3 3 0 0 0 0 0 0 0 0	57.39 49.86 62.25 63.44 43.78 87.4 86.42 71.5 79.55 80.64 37.28 65.18 41.82 41.6 52.74 61.34 41.6 67.96	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2569 2156 9261 24652 3859	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 56.9 26 34.4 43.5	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.9 4.6 2.8 2.8 2.8 2.8 2.4 4 2.6 1.7 7 2.2	92,118 81,295 88,013 95,514 75,041 87,619 84,548 74,841 60,571 85,623 70,923 99,174 48,539 77,320 77,526 68,267
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 57.4 86.42 71.5 79.55 80.64 37.28 65.18 41.82 41.8 52.74 61.34 41.24 61.34 41.24 67.96 32.73	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 0 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2569 2156 9265 24652 3859 2158	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1 9.1	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 56.9 26 34.4 43.5 21.5	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.2 3.2 3.9 4.6 2.8 2.8 2.8 2.8 2.4 4 2.6 1.7 2.2 3.1	92,118 81,295 58,013 95,514 95,514 87,619 84,548 74,841 60,571 55,623 70,923 99,174 48,539 77,320 79,326
0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 1 0	0 0 0 0 0 0 0 4 0 3 3 3 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 86.42 71.5 79.55 80.64 37.28 65.18 41.82 41.6 52.74 61.34 41.82 41.6 52.74 61.34 41.24 67.96 32.73 42.93	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 27122 62325 8392 2569 9261 2156 9261 24652 2158 3859 2158 1606	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1 9.1 8.7	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 9 26 34.4 43.5 21.5 21.5	35 27 47 46 27 33 32 32 39 46 28 28 24 4 26 17 22 3.1 27	92,118 81,295 88,013 95,514 75,041 87,619 84,548 74,841 60,571 85,623 70,923 99,174 48,539 77,320 77,526 68,267
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0	57.39 49.66 62.25 63.44 43.78 86.42 71.5 79.55 80.64 37.28 80.64 41.6 52.74 65.18 41.6 52.74 61.34 41.24 67.96 53.273 42.93	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 0 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 18566 21004 1542 22122 22559 2156 9261 2559 2156 9265 23859 2158 1606 3770	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1 9.1 8.7 9.8	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 56.9 26 34.4 43.5 21.5 41.7 36	35 27 47 46 27 33 32 32 32 39 46 28 24 4 26 17 22 24 4 26 17 22 39	92,118 81,295 58,013 95,514 75,514 75,514 87,519 84,548 74,841 55,623 70,923 99,174 48,539 77,320 79,526 68,267 76,633
	0 0 0 0 0 0 0 4 0 3 3 3 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 86.42 71.5 79.55 80.04 47.5 80.04 41.82 41.6 52.74 41.6 52.74 61.34 41.2 61.34 41.2 61.34 41.2 63.2 73 42.93 32.73 42.93	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2152 8392 2156 9261 24652 23859 2158 1606 37770 6622	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1 9.1 8.7 9.8 10.6 7.8 17.6	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 56.9 26 34.4 43.5 21.5 41.7 36 21.5 41.7 24.8	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.9 4.6 2.8 2.8 2.4 4 2.6 1.7 2.2 3.1 2.7 3.9 3.4	92,118 81,295 88,013 95,514 75,041 87,619 84,548 74,841 40,571 55,623 70,923 99,174 48,539 97,73,20 79,526 48,539 77,526 48,257 76,633 78,930 56,634 102,383
0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		57.30 40.66 40.25 83.44 43.78 77.5 86.42 77.5 79.55 70.55 80.04 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 41.82 51.34 41.82 51.34 41.82 51.34 41.82 51.34 51.35 51	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 18546 221004 1542 22122 2569 2156 8392 2569 2156 24652 3859 2158 1606 3770 6622 3647	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1 9.1 8.7 9.8 10.6 7.8	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 36.6 35.6 56.9 26 34.4 43.5 21.5 41.7 36 24.8 24.8 51.4	35 27 47 46 27 33 32 32 39 46 28 46 28 4 4 26 17 22 31 27 39 34 34	92,118 81,295 58,013 95,554 75,041 87,649 84,546 74,841 46,571 70,523 70,523 99,174 48,539 77,1320 99,174 48,539 77,1320 79,526 68,267 78,633 78,930 56,634
	0 0 0 0 0 0 0 4 0 3 3 3 0 0 0 0 0 0 0 0	57.39 49.66 62.25 63.44 43.78 86.42 71.5 79.55 80.04 47.5 80.04 41.82 41.6 52.74 41.6 52.74 61.34 41.2 61.34 41.2 61.34 41.2 63.2 73 42.93 32.73 42.93	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 1 1 1 1 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0		1 0 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	6291 4234 2276 1091 8566 21004 1542 22122 62325 8392 2152 8392 2156 9261 24652 23859 2158 1606 37770 6622	7.2 9.6 8.2 15.3 16.5 11.9 10.5 15.1 11.9 16.2 17.1 12.9 5.3 22.1 9.1 8.7 9.8 10.6 7.8 17.6	42.3 39.6 27.2 39.6 45.5 55.6 49.9 32.5 41.3 36.6 35.6 56.9 26 34.4 43.5 21.5 41.7 36 21.5 41.7 24.8	3.5 2.7 4.7 4.6 2.7 3.3 3.2 3.9 4.6 2.8 2.8 2.4 4 2.6 1.7 2.2 3.1 2.7 3.9 3.4	92.118 81.295 58.0133 95.514 75.641 78.641 84.548 78.841 46.571 79.923 79.926 68.267 79.526 68.267 76.633 76.930 76.930 56.634

	used# ProtestorsBid						County RedStateRedC					ationUnen	nploymentMedia	In Household IncomePr	
0	0	18.4				0	0 (16.9	20.7	3.7	48,912
0	0	40.0			1 1	1	1 (15.7	24.1	6	60,648
0	0	79.8	13 0) 1	1 1	1	1 ()	0	0 :		9.5	51.5	4.1	121,190
0	0	69.0	17 0) () 1	1	0 0)	0	0			33.9	3.8	58,375
0	0	69.0	17 0) (o 1	1	0 0	0	0	0	3196	17.1	33.9	3.8	58,375
0	0	70.4) 1	1 1	1	1 (0	0		1050	10.1	53.3	2.6	89,418
0	0	50.7	4 0) 1	1 1	1	1 (1	0			13.1	39	2.1	79,357
1	0 2	50.8			1		1 (34.5	2.3	68,279
0	0 2	25.8					0					7.4	29.7	2.1	72,091
0	0	42.2					0 (16.4	25		59,358
														4.3	
0	0	73.					0 (5403		31.1	3.1	47,479
0	0	52.1					0 0					25.5	45.8	3.3	47,284
0	0	48.0					0 (0			26.6	3.3	49,344
0	0	70.2	1 0) () (0	0 0	0	0	0	2580	28.8	20	4.9	35,520
0	0	61.1	7 0) () (0	0 0)	0	0 !	5365	9.7	46.9	3	79,609
0	0	25.7	9 0) () ()	0 0	0	0	0	5711	15.5	25.1	3	53,548
0	0	38.7					0 0				5289	14.1	32.4	2.6	55,751
0	0	28.7					0						31.5	3.1	54,740
0	0	30.4					0				1946	17.6	26	3	
0	0	36.6					0 0						23.3		54,623
														2.9	56,751
0	0	71.					1 (11.9	45.5	2.7	75,041
0	0	54.1			1 1		1 (48.7	3.1	135,528
0	0	3	9 0) 1	1	1	1 (0	0	0	1114	12.1	22	4.2	57,357
0	0	47.9) 1	(1	1 (0	0	0		6.7	50.6	3.8	117,699
0	0	83.2	9 0	1	1	1	1 (0	0			27.7	22	6.8	45,864
0	0	52.1					0 0					8.8	53.1	2.1	83,520
0	0	36.					0 (33.9	2.6	78.216
0	0	41.5				-	0 0				1229		25.6		
0													27.7	2.5	50,518
	0	55.					0							3.4	58,898
0	0	57.5					1 (30.3	5.5	78,779
0	0	59.7					0 0						44.6	2.7	67,906
0	0	74.2	2 0) 1	1	1	1 (0	0		3417		41.9	4.4	76,614
0	0	49.6) (1	0 0	0	0				33.8	3.5	75,468
0	0	38.5					0				0349	24	27		50,498
0	0	72.5					0				2573	13	58	5.2	
0	0						0 0							3.4	89,798
		72.5									1798	13	58	3.4	89,798
0	0	87.2					1 (35.4	2.9	54,735
0	0	41.2					0 0						23.4	2.9	66,830
0	0	30.7					0					13.2	23	4.1	54,612
0	0	64.6		0 0	0 0	0	0	0	0	0		15.1	42	3.1	69,762
0	0	72.1	2 0) 1	1 1	1	1 (0		2883	12.4	51.1	2.9	81.878
0	0	69.0				1	0		0				33.9	3.8	58,375
0	0	55.6					0						34.8		74,134
														3.5	
0	0	57.1					0	,					41.4	3.2	67,033
0	0	49.2			1		1 (0					32.2	3.4	88,532
0	0	71.0	13 0) 1	1 1	1	1 ()	0	0			35.5	5	82,455
0	0	53.3	14 0) 1	1 1	1	1 ()	0	0			44.9	2	116,796
0	0	27.	.3 0) ()	0 0)	0	0	2810	11.4	24.3	3.3	68,334
0	0	53.0	5 0) 1	1	0)	0	0	2870	12.1	33.4		72,590
0	0	33.2				1	0					17.6	32.3	3.6 4.4	51,949
0	0	71.0				1	1 (13.7	35.5	4.4	82,455
0	0	29.5						<u></u>					20.3		58,791
0		77.1						() () () () () () () () () ()					63.9	3.8	96.584
	0						1 (1215	11	34.1	2.8	
0	0	35.8			0 0		0 0)			1788	11.6	41.9	1.9	67,573
0	0	74.2	2 0) 1	1 1	1	1 ()	0	0				4.4	76,614
0	0	60.2	1 0) 1	1 1	1	1 ()	0	0 2	1431 .	13.2	42.1	3.9	98,365
0	0	68.0) 1	1 1	1	1 ()	0		1696	10.1	12.5	5.2	43,720
0	0	69.5		1		1	1 (3784		27.9		91,450
0	0	59.2				1					0000		41.2	7	68,169
													42.2	3.7	74,424
0	0	54.3					0 (1364 .	11.5	41.1	2.6	
0	0	52.3					0 (2574	10.8	41.6	2.1	68,358
0	0	62.7) 1	1	0	0					27.3	3.2	83,856
0	0	53.6	i6 0	0 0)	0)	0				39.7	5.4	70,838
0	0	44.2				0	0 0				1000	12.9	38.1	3	78,309
0	0	53.8				1	1 (9.5	34.3	1.9	84,898
0	0	52.7				1	1 0				2600	7.7	34.3 49.9		84,230
0												9	49.9	2.7	84,230
	0	80.6					1 (1087		47.9 37.9	3.2	
0	0	72.4			1	1	1 (8871 .	15.2		4.4	82,361
0	0	77.0			1 1	1	1 (3482	14.2	26.2	5.5	74,747
0	0	68.4			1 1	1	1 (0	0	0			21	4.6	47,675
0	0	51.9		1	1 1	1	1 (0		2517	25.2	31.4	4.2	42,280
1	3	58.0	3 0) 1	1 1	1	1 (0	0				35.5	3.5	65,034
0	1 0	71.0			1 1	1	1 (21.40	19.8	48.7	5	82,455
0	0	54.1				1	1 0					13.7	39.6	3.1	135,528
0	0	86.4				1	1 0				1220	5.8	52.5		95,514
0											2502		39.6	4.6	
	0	67.5					1 (2593	9.3	23.5	3.4	108,037
1	4	86.4					1 (5559	16.5	31.6	4.6	95,514
0	1 0	37.3					0 (1492		27.2	2.9	55,050
0	0	42.6	i4 0) (1	1	0 (0	0	0		13.6	39	2.9	64,720
0	0	44.2) 1	1 1	1	1 (0		1004		38.1	4.1	63,565
0	0	57.5	8 0) 1	1 1	1	1 (0		1022	13.8	53.5	3.5	86,078
0	0	60.8					0			0 1	14.00	10.9	56.3	3.9	63,822
0	0	80.4					0 0					14.8	21.7	3.1	79,524
1	4	62.2			,) 1		0 (1/70	10.6	51.4	3	97,099
0	4											7.2			
0	0	49.6			1		0 (2102	14.4	53.3	4.6	55,406
0	0	57.6			1 1						3481	6.6	43.3	3.4	102,383
0	0	70.4					1 (1297	10.1	40	2.6	89,418
	0	46.8	14 0) (0 (0	0		200	9.7	41.9	1.7	70,002
0	0	39.9					0 0				00/0		42.1	2.6	71,833
	0	74.2									202		31.5	4.4	76,614
0											2007		41.9		98,365
0 0 0	0	60.2					1 (10.1	29.4	3.9	72,965
0 0 0		42.9					0 (0 :	1418 .	12.9	49.9	3.5	
0 0 0 0	0	74.2									3323		49.9	4.4	76,614
0 0 0 0 0	0		2 0) () (0							39.6	3.5	49,403
0 0 0 0		34.0													04540
0 0 0 0 0	0	34.0) 1	1 1	1	1 ()	0	0 2			32.7	3.2	84,548
0 0 0 0 0 0 0 1	0 0 4	34.0 80.6	14 0								7955	15.1	32.7		
0 0 0 0 0 0 1 0	0 0 4 1 0	34.0 80.6 80.6	4 0 4 0	0 1	1 1	1	1 (0	0	0 1	2592		32.7 39.5	3.2	84,548
0 0 0 0 0 0 1 0 1	0 0 4 1 0 4	34.0 80.6 80.6	4 0 4 0) 1) (1 1 D 1	1	1 () D ()	2 2	0	0 1	7955 2592 3051	15.1 15.1 17.8		3.2 4.2	84,548 64,016
0 0 0 0 0 0 1 0	0 0 4 1 0	34.0 80.6 80.6	4 0 4 0 9 0) 1) ()) 1	1 1 2 1 1 1	1	1 ()))	0 0 0	0 11 0 33 0 14	7955 2592 3051	15.1 15.1		3.2	84,548

ncampmen	tForce used	Protestors P	iden County Vote	Blue CountyRis	ue StateBlue	GovernorBlue	County BlueStateBlue	County RedStatePed	County BlueStateRed	County RedStateUni	PopulationPove	rty RateEduca	tionUn	mploymentMedia	n Household Income	Private Universi	si
1	1 0	2	54	4.5 0	0		1	0	0	0	0 8	388 1	4.3	35.8	4.6	61,598	0
0	0	0					1	1	0	0			8.5 9.6	18.8 43.8	3.4	56,353	0
0		0							0	0			9.6 5.1	43.8	2.5 2.8	65,190 57,507	0
	0	0	28.				0	D	0	0		490 1	0.1	22.8	3.4	70,594	1
	0	0							0	0			0.1	22.8	3.4	70,594	1
0	0	0	74.				1		0	0			8.8 8.8	55.9 55.9	3.4 3.4	116,044	1
	0	0	15.				0	D	0	0			6.9	30.4	1.9	116,044 75,994	1
	0	0	70.	46 C					0	0	0 1	393 1	0.1	53.3	2.6	89,418	1
C		0						0	0	0			18	32.5	1.9	54,020	0
1	1 0 0	0 4	74.				1	1	0	0			3.2 8.6	41.9 47	4.4	76,614	1
	0	0	87.				1		0	0			0.2	35.4	2 2.9	77,432 54,735	— i
C	0	0	64.	48 0	0			0	0	0	0 25	444 1	2.7	35.8	2.9	70,834	1
	0	0						D	0	0	0 1		11	19	2.4	66,247	1
0	0	0							0	0			11 8.3	19 50.2	2.4	66,247 90,801	1
1			47.					0	0	0			0.3	27.9	3.8	67,124	1
	1 0	3	71.	03 0					0	0	0 2		3.7	35.5	5	82,455	1
0		0							0	0			7.1	9.8	5.8 3.3	40,683	1
	0	0	83.					0	0	0			3.3	46.7 21.3	3.3	76,736 75,916	
	0	0							0	0			5.1	42	3.1	69,762	1
c		0							0	0		392 1	4.8	18.9	3.8	58,844	1
0		0	29. 64.					0	0	0		127 1	4.1	19.9 42	3.7 3.1	60,384	1
		3							0	0			5.1 4.4	17.9	4.1	69,762 54,906	0
c	0	0	29.	67 0	0		0	D	0	0	0 1	322	16	13.7	3.4	59,012	C
0		0	35.					0	0	0	0 3	115	9.7	29.3	3.1	76,596	0
	0	0						0	0	0			5.3	35.4 58	4.4 2.8	48,265 121,528	1
		0							0	0		892 097 1	5 7.1	21	2.8	121,528	
c	0	0	16.	19 0			0	0	0	0	0 1		1.1	24.4	2.1	60,719	
	0	0	36.	78 0				D	0	0	0 30	037	22	40.1	3	51,236	0
0		0					-	0	0	0			5.1	30.5 30.4	3.4 3.6	61,704 71,499	1
	0	0					1	1	0	0			0.1	20.4	5.8	70,494	- 1
0	0	0		21 0	1		1	1	0	0			2.8	48.6	3.6	79,432	C
0		0					1	1	0	0	0 5	904 1	9.3	21.5	5.2	54,961	1
1	1 0 0	4	67.				1	1	0	0			6.2	54.4 35.5	3 5	70,117 82,455	0
		0					1	D	0	0		738 1 015	3.7 6	25.3	2.5	67,263	- 1
c	0	0			0		0	D	0	0			5.1	42	3.1	69,762	1
	0	0						D	0	0		083 1	8.9	25.5	4.7	50,519	1
0		0					0	1	0	0			4.7	31.5 39.6	3.7 6.2	65,839	1
	0	0					1	1	0	0			6.5 0.1	42.1	6.2 3.9	84,738 98,365	1
C	0	0	53.	76 0	1		1	1	0	0	0 3	692	9.9	30.1	4.7	92,793	1
0		0					1	1	0	0		561 1	3.7	35.5	5	82,455	1
	0	0					1	1	0	0			8.9 8.1	38.5 46.7	3.5 3.2	98,580 98,706	1
		0					1	1	0	0			8.1 4.9	35.9	4.1	89,334	1
C		0	55.	97 0	0		0	D	0	0	0 4		0.8	39.6	3	76,713	1
C		0					1	1	0	0			7.6	55.9	3.5	150,502	1
0	0	0						0	0	0			6.1	45.6 34.8	2.6 3.8	93,925 70,871	1
0		0							0	0			3.8 0.3	34.6	4.2	56,385	1
0		0					1	0	0	0	0 5		2.8	24.8	3.5	61,729	1
	0	0						D	0	0	0 42		7.1	46.5	2.7 2.8	67,654 106,743	1
0	0	0						0	0	0	0 21	323	6.9	52.3 24.1	3.3	57,884	0
0		0						0	0	0	4		4.4 1.9	27.2	3.5	72,157	0
	0	0						D	0	0	0 2	196 1	0.8	41.6	3.2	83,856	C
0		0							0	0	2	295	4.5	29.6	3.8 3.2	56,489	0
0	0	0	53					0	0	0	0 9	523 1	2.9 8.2	34.3 24	3.2	67,813 73,709	0
C	0	0					1	1	0	0	0 1:	102	3.7	35.5	5	82,455	1
	0	0	26.	84 0					0	0	0		1.8	29.2	2.2	61,933	
0	0	0						0	0	0		411	11	19.7 34.6	3 4.2	58,642 56,385	
0		0							0	0	0 3	610 2	.03	24.2	3.1	62,428	
C		0	36.	96 0	0		1	0	0	0	0 2		3.5 7.3	32	2.8	51,835	
	1 0	3					1		0	0	0		3.7	35.5	5	82,455	1
	0 0	' O							0	0		069 1	3.8	35.3 44.8	4.6 1.9	80,180 81,205	
	0	0						1	0	0	0 5	215	9.4	44.0	6.2	84,738	1
C	0	0	59.	43 0	0		1	D	0	0	0 5		0.1	44.8	3.3	71,973	1
	0	0							0	0	0		8.1	22.5	2.8	55,852 82,455	
	1 0 ⁰	3	71.						0	0	0 2	296 1	3.7	35.5 48.6	3.6	79,432	
	0	0						D	0	0	0 26		2.8	26.6	4.6	75,223	
0	0	0	76.	78 0			1	1	0	0	0		3.9 9.1	41.3	5.5	73,244	1
C		0	33. 34.					0	0	0	0 1		6.7	17.9 35.9	3.2 3.6	53,847 64,187	
0	0	~						0	0	0	0 1	285 1	2.8	35.9 44.9	3.9	94,832	1
0	0	3	69				1	1	0	0	0 14		0.6	32.2	3.3	74,190	1
0 0	0	3		52 0	1		0	D	0	0	0		2.7	31.1	3.1	65.967	0
0 0 1 0 0	0 1 0 1 0	3	60. 43.	16 0	0								13				
	0 1 0 1 0	3 0 0 0	60. 43. 43.	16 C	0		0		0	0	0	389 389	13 13	31.1	3.1	65,967	
	0 1 0 1 0 0	3 0 0 0	60. 43. 43. 56.	16 0 16 0 67 0	000000000000000000000000000000000000000		0	D	0	0	0 9	389 505 1	13 4.3	31.1 24.9			(
	0 1 0 1 0	3 0 0 0	60. 43. 43. 56. 56.	16 0 16 0 67 0 67 0			0	0			0 9 0 10 0 10	389 505 1 505 1	13 4.3 4.3	31.1	3.1 5	65,967 66,169	0
	0 1 0 1 0 0 0 1 1	3 0 0 0 0 0 0	60. 43. 56. 56. 48. 66.	16 00 16 00 67 00 67 00 75 00 68 00			0	0	0	0	0 9 0 10 0 10 0 58	389 505 1 505 1 024 1	13 4.3 4.3 7.1	31.1 24.9 24.9 40.1 48.6	3.1 5 5 2.9 3.4	65,967 66,169 66,169 56,088 80,645	
	0 1 0 1 0 0 0 1 1 0 0		60. 43. 43. 56. 56. 48. 66. 25.	16 00 16 00 67 00 67 00 75 00 68 00 75 00			0 0 0 1	D D D 1	0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 9 0 10 0 10 0 58 0 2 0 2	389 505 1 505 1 024 1 881 1	13 4.3 4.3	31.1 24.9 24.9 40.1 48.6 27	3.1 5 5 2.9 3.4 3.6	65,967 66,169 66,169 56,088 80,645 61,747	
	0 1 0 1 0 0 0 1 1 0 0		60. 43. 43. 56. 56. 48. 66. 25. 58.	16 00 16 00 67 00 67 00 68 00 75 00 68 00 75 00 05 00	0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1		0 0 0 1 1 1	D D D 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 9 0 10 0 10 0 58 0 2 0 1 0 1	389 505 1 505 1 024 1 881 1 498 1	13 4.3 4.3 7.1 0.2	31.1 24.9 24.9 40.1 48.6 27 41.1	3.1 5 5 2.9 3.4 3.6 3.9	65,967 66,169 66,169 56,088 80,645	0 0 1 1 1
	0 1 0 1 0 0 0 1 1 0 0		60. 43. 43. 56. 56. 48. 66. 25. 58. 57.	16 0 16 0 67 0 67 0 68 0 75 0 05 0 44 0	0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 1		0 0 0 0 1 1 1 1 1 1	D D D 1 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 9 0 10 0 10 0 58 0 2 0 11 0 11 0 6	389 505 1 505 1 024 1 881 1 498 1 564 1 947	13 4.3 7.1 0.2 2.2 2.5 6.6	31.1 24.9 24.9 40.1 48.6 27	3.1 5 5 2.9 3.4 3.6	65,967 66,169 56,088 80,645 61,747 84,615 84,738 54,517	0 0 1 1
	0 1 0 1 0 0 0 1 1 0 0 0 0 0 0 0 0		60. 43. 43. 56. 56. 48. 66. 25. 58. 57. 36. 79.	16 0 16 0 67 0 67 0 68 0 75 0 05 0 57 0 57 0 57 0 21 0			0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	D D D 1 1 1 1 1 1 1 1	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 9 0 10 0 10 0 58 0 2 0 11 0 11 0 6 0 10	389 505 1 505 1 024 1 881 1 498 1 564 1 947 519	13 4.3 7.1 0.2 2.2 2.5	31.1 24.9 24.9 40.1 48.6 27 41.1 52.6	3.1 5 5 2.9 3.4 3.6 3.9 6.2	65,967 66,169 56,088 80,645 61,747 84,615 84,738	1

nomentForce use	d# ProtestorsBiden (ounty VoteBlue County	Blue StateBlue Gove	morBlueCounty Blu	eStateBlueCounty Re	StateRedCounty Blue	StateRedCounty RedS	tatelini Population	Poverty Rate	Education	nemploymen	tMedian Household In	comePrivate Universi
0	0	79.55	0 1	1	1	0	0	0	7538	10.5	55.6	3.3	87,619
0	0	29.53	0 0	0	0	0	0	0	1704	6.6	39.5	2.6	99,932
0	0	86.42	0 1	1	1	0	0	0	4199	16.5	39.6	4.6	95,514
0	0	51.59	0 1	1	1	0	0	0	9174	12.4	37.3	3.3	80,040
0	0	60.52 60.52	0 1	1	1	0	0	0	8190	12.7 12.7	32.2 32.2	3.3	74,190
0	0	64.42	0 1	0	0	0	0	0	3428 2411	12.7	34.2	3.3 4.3	74,190 61,452
	0 3	55.94	0 0	0	0	0	0	0	11186	16	33.7	4.3	68,748
0	0	69.14	0 1	1	1	0	0	0	5766	10.6	44.9	3.9	94,832
0	0	44.29	0 0	0	0	0	0	0	2055	9.5	39.7	3	78,309
0	0	52.81	0 1	0	1	0	0	0	2552	6.5	40.6	2.3	96,415
0	0	59.43	0 0	1	0	0	0	0	4867	11.6	44.8	3.3	71,973
0	0	59.25	0 1	1	1	0	0	0	1995	13.1	41.2	3.7	68,169
0	0	59.25	0 1	1	1	0	0	0	20200	13.1	41.2	3.7	68,169
0	0	56.24	0 0	1	0	0	0	0	1471	8.3	50.2	2.9	90,801
0	0	86.42 49.98	0 1 0	1	1	0	0	0	1721 1563	16.5 14.7	39.6 24.6	4.6	95,514 62,557
0	0	59.82	0 0	0	0	0	0	0	4159	14.7	33.5	3.4	62,509
0	0	59.82	0 0	0	0	0	0	0	4159	12	33.5	3.4	62,509
0	0	36.6	0 0	0	0	0	0	0	1273	9.9	33.9	2.6	78,216
0	0	57.88	0 1	1	1	0	0	0	2046	7.4	50.7	2.9	102,413
0	0	26.3	0 0	0	0	0	0	0	1388	9	36.9	2.6	95,085
0	0	45.19	0 1	1	1	0	0	0	1778	3.7	60.9	2.9	140,768
0	0	63.52	0 1	1	1	0	0	0	5862	7.9	51.8	2.5	101,158
0	0	21.52	0 0	0	0	0	0	0	3809	9.6	25.6	6.2	84,738
0	0	60.85	0 0	0	0	0	0	0	3747	12.4	38.4	2.8	72,129
0	0	74.22	0 1	1	1	0	0	0	5724	13.2	41.9	6.2	84,738
0	0	60.78 41.47	0 1	1	1	0	0	0	2952	8.2 20.2	47 25.9	5	101,621
0	0	41.47 53.66	0 0	1	0	0	0	0	2552 2044	20.2	25.9	3.8	52,286
0	0	62.41	0 0	1	0	0	0	0	2044	6.9	52.3	5.4	70,838
0	0	49.99	0 1	1	1	0	0	0	23638	8.4	35.8	2.8 4.3	106,743
0	0	74.22	0 1	1	1	ő	0	0	5365	13.2	41.9	4.3	97,851 76,614
0	0	51.59	0 1	1	1	0	0	0	2752	12.4	37.3	3.3	80,040
0	0	65.91	0 1	1	1	0	0	0	8332	12.3	35.3	4.8	81,372
1	0 4	60.22	0 1	1	1	0	0	0	69755	8.6	45.1	4.0	102,073
0	0	77.07	0 1	1	1	0	0	0	15191	14.2	37.9	5.5	74,747
0	0	62.9	0 1	1	1	0	0	0	10896	11.3	41.7	3.9	81,042
0	0	52.81	0 1	0	1	0	0	0	2690	6.5	40.6	2.3	96,415
0	0	62.25	0 0	1	0	0	0	0	1405	7.2	56.3	3	97,099
0	0	37.58 71.41	0 1 0	0	0	0	0	0	13081	11	28.6 55.5	2.8	72,378
0	0	51.14	0 1	0	1	0	0	0	4300 1580	10.2	57.2	3.3	95,151 131,562
0	0	30.71	0 0	1	0	0	0	0	3330	4.8	23	3.7	54,612
0	0	59.25	0 1	1	1	0	0	0	4418	13.2	41.2	4.1 3.7	68,169
0	0	37.58	0 1	1	1	0	0	0	2134	13.1	28.6	2.8	72,378
0	0	66.45	0 1	1	1	0	0	0	2434	6.9	52.2	2.4	88,571
0	0	81.21	0 0	1	0	0	0	0	7940	20.3	34.6	4.2	56,385
0	0	39.35	0 0	0	0	0	0	0	11331	10.5	28	3.3	65,933
0	0	81.93	0 0	0	0	0	0	0	17516	20.1	40.2	3.7	52,278
0	0	81.93	0 0	0	0	0	0	0	17516	20.1	40.2	3.7	52,278
0	0	57.46	0 1	1	1	0	0	0	2014	9.9	37.2	4.1	88,560
0	0	41.47	0 0	0	0	0	0	0	1407	20.2	25.9 32.6	3.8	52,286
0	0	51.98 71.63	0 1	1	0	0	0	0	2054 4655	14.2	44.9	3.9	61,028
0	0	49.07	0 1	1	1	0	0	0	4655 6015	8.3	32.1	4.1	119,667 65,699
0	0	75.78	0 1	0	1	0	0	0	2143	12.1	55.8	2.3	86,579
0	0	45.49	0 0	1	0	0	0	0	2594	7.8	32.4	1.6 2.7	72,398
0	0	72.45	0 1	1	1	0	0	0	3707	9.7	47.9	4.4	82.361
0	0	35.16	0 0	1	0	0	0	0	2088	15.2 10.8	31.8	3.5	71,152
0	0	74.22	0 1	1	1	0	0	0	4411	13.2	41.9	4.4	76.614
0	0	63.44	0 1	1	1	0	0	0	8518	9.6	42.3	3.5	92,118
0	0	30.25	0 0	0	0	0	0	0	1168	14.4	25.9	3.3	57,764
0	0	47.72	0 1	1	1	0	0	0	9718	15.4	29.1	2.5	68,851
0	0	62.41	0 0	1	0	0	0	0	1624	6.9	52.3	2.8	106,743
0	0	63.89	0 1	1	1	0	0	0	3331	11.3	53	2.6	92,795
0	0	33.39	0 0	0	0	0	0	0	24407	22.4	21.4	4.6	53,301
0	0	79.83 60.21	0 1	1	1	0	0	0	2569	9.5	51.5 42.1	4.1	121,190
1	0	60.21 85.26	0 1 1	1	1	0	0	0	40053 30596	10.1	42.1	3.9	98,365 135,366
1	0 3	72.64	0 1	1	1	0	0	0	30596	11.9	55.9	3.3	135,366 150,502
0	0 0	72.64	0 1	1	1	0	0	0	10459	7.6	55.9	3.5 3.5	150,502
0	0	67.57	0 1	1	1	0	0	0	2037	7.6 9.3	52.5	3.5	108,037
0	0	58.62	0 0	0	0	0	0	0	16051	9.3	36.8	2.9	64,029
0	0	58.62	0 0	0	0	0	0	0	4001	15.2	36.8	2.9	64,029
0	0	71.03	0 1	1	1	0	0	0	1036	13.7	35.5	5	82,455
1	3	74.22	0 1	1	1	0	0	0	4179	13.2	41.9	4.4	76,614
0	1 0	86.42	0 1	1	1	0	0	0	5247	16.5	39.6	4.6	95,514
0	0	23.51	0 0	0	0	0	0	0	1502	12	31.2	3.5	59,619
0	0	71.03	0 1	1	1	0	0	0	1344	13.7	35.5	5	82,455
0	0	74.95 74.95	0 1	1	1	0	0	0	4452 8581	8.8	55.9 55.9	3.4	116,044
0	0	74.95	0 1	1	1	0	0	0	11614	8.8	37.9	3.4	116,044 74,747
0	0	35.16	0 0	1	0	0	0	0	2366	14.2	31.8	5.5	74,747 71,152
0	0	62.25	0 0	1	o	ő	0	0	1609	10.8	56.3	3.5	97,099
0	0	28.33	0 0	0	o	o	0	0	4081	7.2	18	4.8	48,465
	0	43.77	0 0	0	0	0	0	0	3665	18.6	33.7	4.8	87,259
0	0	43.78	0 0	1	0	0	0	0	6921	14.8	39.6	2.7	81,295
0	0	28.81	0 0	0	0	0	0	0	1670	0.2	24.9	3.2	56,651
0		64.55	0 1	1	1	0	0	0	4252	12.9	45.2	3.3	76,997
0 0 0 0 0	0	39.13	0 0	1	0	0	0	0	2437	10.6	22.3	4.2	67,486
0 0 0 0	0		0 1	1	1	0	0	0	7499	15.1	49.9	3.2	84,548
0 0 0 0 0	0	80.64			0	0	0						
0 0 0 0 0 0	0	80.64 40.51	0 0	0				0	1605	5.8	32.7	2.6	90,943
0 0 0 0 0 0 0	0 0 0 0 0	80.64 40.51 32.28	0 1	1	1	0	0	0	1083	5.8	23.3	2.6 5.1	66,312
0 0 0 0 0 0 0 0 0 0	0 0 0 0	80.64 40.51 32.28 51.62	0 1 0 1	1	1	0	0	0	1083 3596		23.3 43.8	5.1 2.9	66,312 93,341
0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0	80.64 40.51 32.28 51.62 33	0 1 0 1 0 0	1 1 1	1 1 0	0 0	0 0 0	0 0 0	1083 3596 9801	5.8 13.2 7.2 9.3	23.3 43.8 38.6	5.1 2.9 3.1	66,312 93,341 81,507
0 0 0 0 0 0 0 0 0 0	0 0 0 0	80.64 40.51 32.28 51.62	0 1 0 1	1	1	0	0	0	1083 3596	5.8 13.2 7.2	23.3 43.8	5.1 2.9	66,312 93,341

Encampmer	tForce used#	ProtestorsBiden Co	unty VoteBlue CountyB	lue StateBlue	GovernorBlueC	ounty BlueStateBlue	County RedStateRed	County BlueStateRedC	ounty RedStateUni	PopulationPover	v RateEducatio	Unemploym	entMedian Househo	old IncomePrivate Unive	rsi
(0	0	66.24	0 0	0	()	0	0	0 29	08 21.7	19.4	4.4	42,209	(
	0	0		0 0					0	0 62	67 13.8		2.8	70,013	1
	0	0		0 0					0	0 29			1.9 3.6	67,573 48,792	
	0	0		0 0					0	0 12			4.3	68,748	
	0	0	52.71	0 0					0	0 10	13 13.2	37.2	2.9	74,091	1
	0	0		0 0					0	0 112	05 12.6		2.7	64,118	(
	0	0		0 0		(0	0 37			3	97,099	1
	0	0		0 0					0	0 61 0 104		24.4	2.9 3.7	53,782 61.941	- 1
	0	0		0 0					0	0 32			3.2	75,127	
	0	0	32.66	0 0	0	()	0	0	0 48			4.3	47,278	(
	0	0		0 1	1	1			0	0 14			5	82,455	1
	0	0		0 1	1	1			0	0 110		41.1 35.1	3.9 4.4	84,615	0
	1 0 0	2		0 1	1				0	0 161 0 153			4.4	44,756 67,441	
	0	0		0 0	0	(0	0 152			3.8	70,871	1
	0	0	53.85	0 1	0			0	0	0 1446	45 7.7	38.1	1.9	84,898	1
	0	0		0 1	1	1			0	0 58			4.3	68,239	(
	0	0		0 0					0	0 18			2.8 3.2	72,129	
	0	0		0 0					0	0 12 0 137			2.6	71,607 66,499	
	0	0		0 0					0	0 16			2.9	55,333	
	0	0	19.11	0 0	0	()	0	0	0 35	48 14.6	21	3.1	53,693	
	0	0		0 1	1	1			0	0 76	98 11.2		2.5	70,958	
	0	0		0 0		(0	0 22			3.7 3.7	89,314	
	0	0		0 0	1	(0	0 29 0 189			3.9	76,285 98,365	
	0	0		0 0	1				0	0 109			3	56,416	
	0	0		0 1	1				0	0 11		35.5	5	82,455	
	0	0	22.16	0 0					0	0 55	23 18.2		2.8 3.4	54,056	
	0	0		0 0		(0	0 19			3.4	101,891 64,719	
	0	0		0 0					0	0 19			3.9	64,719	
	0	0		0 0					0	0 19			3.4	89,798	
	0	0	39.49	0 0)	0	0	0 36	28 13.7	23	4	61,299	
	0	0		0 1	1				0	0 38	20 17.6	28.9	4.3 4.3	64,030	
	0	0		0 1	1				0	0 31 0 51			2.7	52,383 75,041	
	0	0		0 1	1				0	0 51 0 33			5.5	73,244	
	0	0		0 1	1			0	0	0 30			4.4	59,451	
	0	0	48.94	0 1	1			0	0	0 37		29.7	2.6	77,962	
	0	0		0 0	0	()		0	0 290	46 11.3	35.8	2.8 6.3	66,427	
	0	0		0 0	1	()		0	0 11 0 226			4.6	45,996 80.180	
	0	0		0 1	1				0			39.9	3.4	119,253	
	0	0		0 1	1				0	0 58			5.5	73,244	
	0	0	41.57	0 1	1			0	0	0 19		35.1	2	112,154	0
	0	0		0 0	0)		0	0 41	70 14.7	31.5	3.7	65,839	1
	0	0		0 1	0	1			0	0 21			1.8	99,839 66,952	
	1	3		0 1	1	1			0	65			3.5	150,502	
	1 1	3		0 1	1			0	0	0 326		37.1	3.4	79,340	
	0	0		0 0		()	0	0	0 143		26.5	4.3 3.3	51,708	
	0	0		0 0	0			0	0	0 54	75 11.9		4.4	64,601 82,361	-
	0	0		0 1	1				0	0 84	11 15.2		2.9	54,735	
	0	0		0 1	1				0	42			5.9	76,108	
	0	0		0 1	1			0	0	0 114 0 32		30.6	4.1	80,702	0
	1 1	3		0 1	1	1			0	0 321		39.9	3.4 4.9	119,253	
	0 '	0		0 1	1	1			0	0 31	48 15.2		2.9	99,897 74,091	
	0	0		0 0					0	0 21			3.3	76,736	
	0	0		0 1	1				0	74		36.1	2.2	93,833	
	0	0	51.53	0 0	1			0	0	0 21		43.9	3	105,202	
	0	0		0 0					0	0 28		40	2.6 4.3	71,833	
	0	0		0 0					0	0 41	02 18.2	34.2	4.5	61,452 95,151	
	0	0		0 0		1			0	0 34	01 10.2	55.5 49.9	3.2	95,151 84,548	
	0	0		0 0					0	0 81 27		45.3	3.2	52,388	
	0	0	58.87	0 0	1			0	0	0 38		36.6	3.9 3.9	64,719	
	0	0		0 0	1				0	0 38		30.0	3.9	64,719 76,997	
	0	0		0 1	1	1			0	0 207	64 12.9		4.3	76,997	
	1	0 3		0 1	1	1			0	0 48	B3 16.7	52.5	3.4	108,037	
	0 1	0		0 1	1	1			0	46		41.2	3.7	68,169	
	0	0	56.46	0 1	1	1		0	0	0 99		36.9	3.8 3.5	69,888	
	0	0		0 1	1				0	0 57		30.0	3.5	75,077	
	0	0		0 1	1	1			0	0 57	16 5.8	40.7	4.3	135,528 59,335	
	0	0		0 1	1	1			0	00		26.8	3.5	65,272	
	0	0		0 1	1				0	0 60 0 37		24.6	4.4	59,451	
(0	0	34.02	0 1	1	1		0	0	0 25		25	4.2	67,975	
	0	0		0 1	1	1			0	0 25		30.1	3.5 4.3	70,939	
	0	0		0 1	1				0	0 43	03 15	24.0	4.5	55,466 59,451	
	0	0		0 1	1				0	0 36			4	61,424	
	0	0		0 1	1	1			0	0 36		28.3	4.2	58,274	
(0	0	76.78	0 1	1	1		0	0	0 38		41.3	5.5	73,244	
	0	0	51.62	0 1	1	1	I	0	0	0 118		43.8	2.9 6.8	93,341 45,864	
	0	0		0 1		1			0	0 20	09 27.7	22 28	3.7	45,864 69,851	
	0	0		0 1		1			0	0 30	21 11	28 33.9	3.8	68,962	
	0	0		0 1		1			0		70 14.2	27.8	3.7	63,894	
		0	25.6	0 0		()	0	0	0 35	10.7	21.4	4.2	56,385	
0	0	0				(0	0			41.6	3.2	83,856	
0	0	3		0 0									3.6		
0	0 1 1 0	3	58.88	0 1	1	1	1	0	0	0 269	66 10.8	38.1	3.5 3.5	70,939	
0 0 0 0 0	0	3	58.88 17.28		1	1)	0		24	66 10.8 00 13.6 01 15.8	38.1 32.5	3.5 3.5 3.5		

Encampmen	Force used#	ProtestorsBiden Co	unty VoteBlue Count	Blue State	lue Governor	BlueCounty BlueState	BlueCounty RedState	RedCounty BlueSta	teRedCounty RedSt	atelloi Population	Poverty Rat	eEducationU	nemploymen	Median Household In	comePrivate Universi	_
Circampinen		0	29.43	0	0	0	0	0	0	0	2610	20	20.4	3.5	51,053	1
C		0	86.42	0	1	1	1	0	0	0	5922	16.5	39.6	4.6	95,514	1
0		0	55.61	0	1	1	1	0	0	0	1023	12.6	21.6	6.2	84,738	1
0		0	81.21 27.34	0	0	1	0	0	0	0	46015	20.3 17.3	34.6 29.2	4.2	56,385 54,996	0
0		0	19.09	0	0	0	0	0	0	0	11533 1278	14.9	17.5	4.2	60,950	1
0		0	61.05	0	0	0	0	0	0	0	9351	22.5	20.9	4	56,289	1
C	0	0	41.43	0	0	0	0	0	0	0	81678	23.7	42.5	3.1	60,355	1
c		0	23.07	0	0	0	0	0	0	0	13510	12.7	21.9	4.1	68,019	1
C		0	47.85	0	0	0	0	0	0	0	12293	17.3	23.7	4.2	60,250	1
0		0	48.56 44.65	0	0	0	0	0	0	0	7900	22.1 11.5	23.7 27.2	4.6	51,342	1
		0	44.65 58.2	0	0	0	0	0	0	0	2670 7479	11.5	31.5	4.5	62,412 65,839	0
		0	28.09	0	0	0	0	0	0	0	2506	14.7	22.1	4.2	52,460	0
		0	49.31	0	0	0	0	0	0	0	14068	11	34.5	3.7	76,285	1
c	0	0	37.04	0	0	0	0	0	0	0	1832	8.1	30.9	3.4	86,043	1
c		0	55.94	0	0	0	0	0	0	0	8357	16	33.7	4.3	68,748	0
C		0	56.04	0	0	0	0	0	0	0	9341	23.5	20.6	5.4	49,583	1
0		0	54.41 33.12	0	0	0	0	0	0	0	41737	9.8	41.9 33.3	3.3 3.3	89,074 59,138	0
		0	33.12	0	0	0	0	0	0	0	45569 10184	17.1	33.3	3.3	59,138	0
		0	66.66	0	0	0	0	0	0	0	2547	18.5	25.4	4.4	53,441	1
c)	0	49.31	0	0	0	0	0	0	0	2769	11	34.5	3.7	76,285	1
C		0	45.15	0	0	0	0	0	0	0	18115	6.1	48.4	3.5	102,711	= 1
C		0	92.15	0	1	1	1	0	0	0	7186	15.2	63.6	4.9	99,897	1
0		0	74.22	0	1	1	1	0	0	0	1386	13.2	41.9	4.4	76,614	1
0		0	71.03	0	1	0	1	0	0	0	3186	13.7	35.5 22.8	5 3.4	82,455 70,594	1
		0	28.61 28.61	0	0	0	0	0	0	0	1414 1414	10.1	22.8	3.4	70,594	- 1
		0	69.14	0	1	1	1	0	0	0	9439	10.1 10.6	44.9	3.9	94,832	0
c	0	ő	64.55	0	1	1	1	0	ő	0	4522	12.9	45.2	3.3	76,997	- 1
c		0	56.64	0	1	1	1	0	0	0	4428	13.3	32.5	3.2	63,141	1
c		0	30.69	0	0	0	0	0	0	0	2651	9.2	24.7	2.8	70,121	1
1		4	57.46	0	1	1	1	0	0	0	2941	9.9	37.2	4.1 4.6	88,560 95,514	0
0		0	86.42 71.03	0	1	1	1	0	0	0	1663 2878	16.5	39.6 35.5	4.6	95,514 82,455	1
		0	80.64	0	1	1	1	0	0	0	2878 1296	13.7 15.1	35.5 49.9	3.2	84,548	- 1
1		3	86.42	0	i	1	i	0	0	0	1296	15.1 16.5	39.6	4.6	95,514	- 1
0		0	51.42	0	0	1	0	0	0	0	112931	17.1	46.5	2.7	67,654	- 1
0		0	58.87	0	0	1	0	0	0	0	4949	16.2	36.6	3.9	64,719	1
c		0	58.87	0	0	1	0	0	0	0	4949	16.2	36.6	3.9 2.9	64,719 72,658	- 1
0		0	30.32	0	0	0	0	0	0	0	5548	10.6	29.7	2.9	68,210	1
0		0	60.64 28.4	0	0	0	0	0	0	0	12017	11.4	45.9 35.5	2.4	57,497	0
		0	52.71	0	0	0	0	0	0	0	1563 10960	12.9	37.2	2.9	74,091	1
		0	64.42	0	0	ő	0	0	0	ő	8256	13.2 18.2	34.2	4.3	61,452	0
c	0	0	44.14	0	0	0	0	0	0	0	13754	12.9	37	3.2	75,127	0
1		3	41.45	0	0	0	0	0	0	0	39517	13.8	40.2	2.8	70,013	0
C		0	21.99	0	0	0	0	0	0	0	8082	18.6	21.6	3.6 3.7	49,815 76,285	0
1		2	49.31	0	0	0	0	0	0	0	52044	11	34.5 55.5	3.3	95,151	0
1		0	71.41 64.89	0	0	0	0	0	0	0	67083 32772	10.2	34.8	3.8	70,871	0
0		0	66.66	0	0	0	0	0	0	0	27843	13.8	25.4	4.4	53,441	0
0		0	58.2	0	0	0	0	0	0	0	38396	18.5	31.5	3.7	65,839	0
0)	0	29.52	0	0	0	0	0	0	0	11071	14.7 13.2	27.7	3.6	67,978	0
C	0	0	55.94	0	0	0	0	0	0	0	16413	16	33.7	4.3 3.7	68,748 65,839	0
C		0	58.2	0	0	0	0	0	0	0	9476	14.7	31.5	4.3	75,947	0
0		0	37.95	0	0	0	0	0	0	0	5759	11.5	34.3	3.3	64,868	0
0		0	25.49	0	0	0	0	0	0	0	6536	11.7	16.6	6.1	48.825	0
0		0	58.04 81.21	0	0	1	0	0	0	0	36826 2074	26.9	20.3 34.6	4.2	56,385	1
0		o	25.42	0	0	ò	o	ő	o	ő	2593	20.3	23.1	3.5	61,417	1
0	0	0	41.51	0	0	0	0	0	0	0	14804	12.9 14	28.9	3.1	61,634	1
C		0	36.25	0	0	1	0	0	0	0	1003	12.6	25.1	3.8	57,702	1
0		0	48.57	0	1	1	1	0	0	0	1912	11.5	30.9	2.6 3.9	63,191 94,832	
0		0	69.14	0	1	1	1	0	0	0	10804	10.6	44.9	4.2	94,832 56,385	0
0		0	81.21 39.26	0	0	1	0	0	0	0	13939	20.3	34.6 37.6	3.7	77,284	1
0		0	39.26	0	0	0	0	0	0	0	2351 1502	10.6	23.6	3.5	52,236	1
c		0	31.98	0	0	0	0	0	0	0	3375	18.4	20.7	3.6	63,221	1
C	0	0	20.08	0	0	0	0	0	0	0	1943	12.8 15.6	14.9	3.6	51,980 80,641	1
0		0	43.36	0	0	0	0	0	0	0	1120	10	48.9	4.6	80,641 95,514	1
0		0	86.42	0	1	1	1	0	0	0	14139	16.5	39.6	4.7	92,711	1
0		0	63.65 53.66	0	1	1	1	0	0	0	1719	10.3	28.4 27.3	5.4	70,838	1
		0	53.66 62.28	0	0	1	0	0	0	0	1986 25129	12.9	41.4	2.9	54,735	0
0		0	59.25	0	0	1	0	0	0	0	25129 1275	9.8	41.4	3.4	63,667	1
0		0	27.89	0	0	0	0	0	0	0	4155	15.7 10.2	24.7	2.7	73,250	1
C	0	0	43.16	0	0	0	0	0	0	0	1041	10.2	31.1	3.1 3.1	65,967 65,967	1
C		0	43.16	0	0	0	0	0	0	0	1041	13	31.1	3.1	65,967 76,614	1
C		0	74.22	0	1	1	1	0	0	0	1390	13.2	41.9	4.4	84,551	1
1		2	63.06	0	1	1	1	0	0	0	2921	11.3	41.7	5	101,621	1
0	,	0	60.78 58.2	0	0	1	1	0	0	0	1918	8.2	47 31.5	3.7	65,839	- 1
		0	92.15	0	1	1	1	0	0	0	3581 2179	14.7	63.6	4.9	99,897	- 1
0		0	16.26	0	0	0	0	0	0	0	3226	15.2	24.3	2.3	62,100	1
C	0	0	35.89	0	0	0	0	0	0	0	5339	10.8 18.2	35	3.6	46,433 46,433	0
	0	0	35.89	0	0	0	0	0	0	0	5339	18.2 18.2	35	3.6 2.9	46,433 118,494	0
c	1	3	71.47	0	1	1	1	0	0	0	17113	18.2	59	4.2	50,410	1
C 1	0 0	0	18.51	0	0	0	0	0	0	0	1940	14.8	17.7	3.5	74,206	1
0 1 0		0	56.58	0	1	0	1	0	0	0	2907	14	35.1	2.1	68,358	1
0 1 0 0	0		52.34	0	0	0	0	0	0	0	1015	10.8	41.1 13.5	5.8	36,573	1
0 1 0 0 0))	0			0						1416	26.1		3.2	67,033	1
0 1 0 0 0 0)))	0	15.93 57.15		0	0				0	1106		41.4	2.2	FR 000	
0 1 0 0 0))))		15.93 57.15 42.82	0	0	0	0	0	0	0	1106 3713	13.6	41.4 26	3.3	57,070	1
	0 0 0 0 0	0	57.15	0							3713	13.6 18.6		3.3	82,248	1
	D D D D D D D	0 0 0 0	57.15 42.82 42.75 56.92	0 0 0 0 0	0 1 1	0 1 1	0 1 1	0 0 0	0 0 0	0 0 0	3713 4673 1562	13.6 18.6 7.3	26 41.1 34.6			0
	2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	0 0 0	57.15 42.82 42.75	0 0 0	0	0	0	0	0	0	3713 4673	13.6 18.6	26 41.1	3.3 3.6	82,248 80,159	1

0		ProtestorsBi			ue StateBlu	e GovernorBlue	County BlueStateBlue	County RedStateRed	County BlueStateRed	ounty RedStateOni_i	FopulationFo	verty RateEd	ucationUr	remploym	entMedian Househol	d IncomePrivate Univer	
		0	55	.82	0	1	1		0	0	0	5517	6.3	45.2	1.8	112,525	
0		0			0	1	1	1	0	0	0	2286	10.1 19.1	42.1 41.3	3.9	98,365	
0						1	1		0		0	3361		41.3 51.5	5.5	73,244	
0		0			0	1	1		0	0	0	3782 1784	9.5 12.7	34.8	4.1	121,190 63,619	
1	0	4				1	1		0	0		37642	14.1	36.9	3.8	69,888	
0		0							0	0		18608	12.9	35.3	3.7	66,034	
1	1	4						0	0	0		58335	14	35.9	3.8	64,007	
0		0	46	.49	0			D	0	0	0	32804	13.5	34.6	2.3	62,100	
0		0							0	0		10287	15.7	37.4	3.2	57,057	
0		0							0	0	0	3059	20.6	23.3	4.2	48,635	
0		0							0	0	0	7949	15.7	37.4	3.2	57,057	
0		0					0		0	0	0	6533	12.8	24.4	3.4	52,679	
0		0				1	1		0	0	0	4930	20.2	35.4	2.9	54,735	
0		0			0	1	1		0	0	0	4811 54449	11.3 9.5	41.7 51.5	3.9	81,042	
1	0	3				1	1		0	0		52829	15.3	44.1	4.1	121,190 82,359	
0	0	0			0	1	1		0	0	0	1374	11.9	60.1	3.3	135,366	
1	1	4			0	1	1		0	0		45633	9.2	43.4	3.6	106,047	
1	1	5			0	1	1		0	0		69845	13.7	35.5	5	82,455	
1	1	3				1	1	1	0	0		10802	18.6	14.8	9	65,253	
1	0	4			0	1	1	1	0	0		30913	11.3	25.1	4.8	86,350	
1	1	4	60	.21	0	1	1	1	0	0		55842	10.1	42.1	3.9	98,365	
1	1	3			0	1	1	1	0	0		16914	11.9	60.1	3.3	135,366	
1	1	3	64	.52	0	1	1	1	0	0	0	31335	14.9	35.9	4.1	89,334	
1	1	4			0	1	1		0	0		23091	12.6	43.6	5.7	99,256	
0		0						D	0	0		11890	12.9	33.2	2.9	63,429	
1	0	2							0	0		77982	12.4	38.4	2.8	72,129	
0		0				-			0	0		11236	13.6	29.1	3.3	60,086	
0		0					0		0	0	0	3427	15.6	29.3 41.9	3.6	55,160	
1	1	4			0				0	0		30198	13.2 13.6	41.9	4.4	76,614	
0		0			0	1	1		0	0		48594		41.4 63.9	3.2	67,033 96,584	
0		0			0		1		0	0		46885 13971	11	41.1	2.8	96,584 82,248	
1		3				1	1	1	0	0		13971 37449	7.3	41.1	3.3	84,551	
0	1	0			0	1	1	1	0	0	0	1683	9.8	41.5	3.3 3.8	84,551	
0		0				0	0	0	0	0	0	2962	11.3 13.8	34.8	3.8	70,871	
ő		0						0	0	0		14723	13.8	30.5	3.8	62,794	
0		0				1	1	1	0	0		28522	10.9	39.2	3.9	85,427	
1	0	3				1	1	1	0	0	0	17194	10.5	55.6	3.3	87,619	
0	0	0	47	.58	0	0	0	0	0	0	0	2711	9.6	32.9	2.9	73,492	
0		0						D	0	0	0	2905	13.8	27.1	3.2	57,997	
0		0						D	0	0	0	69252	18.5	47.7	3.1	58,783	
1	1	2				0	0	0	0	0		50880	24.1	48.7	3.3	49,832	
0		0			0	1	1	1	0	0	0	7965	11.3	41.7 30.9	3.8	84,551	
0		0			0	1	1	1	0	0	0	3729	16.5	30.9	3	71,143 96,304	
0		0			0	1	1	1	0	0		22457	9.1	37.7	2.7	96,304	
0		0					1	1	0	0	0	3526	9.1	40.2	2.7	52,278	
0		0						0	0	0	0	1273	20.1	40.2	3.7	52,278	
0		0							0	0	0	1273	20.1	33.7	3.7 4.3	68,748	
0		0						0	0	0	0	52610 9873	16	33.7	4.3	68,748	
0		0						0	0	0		9673	16	33.7	4.3	84,738	
ő		0						0	0	0	0	5416	16	19.7	4.1	67,905	
0		0							0	0		13185	14.1 14.3	44.9 41.9	2.7	60,108	
0		0			0	1	1	1	0	0		47263		35.1	4.4	76,614	
0		0				1	1	1	0	0	0	5004	13.2 11.9	46.6	4.3	73,684	
1		3				1	1	1	0	0		65110	16.2	34.1	4.1	61,205	
0	0	0	63	.35	0	0	0	0	0	0	0	6647	15.6	34.1	3.3	62,776	
0		0			0	0	0	D	0	0	0	6647	15.6	54.6	3.3	62,776 74,909	
1		2	70	.57	0	0	0	0	0	0	0	41273	15.5	24.4	2.4	65,633	
0		0							0	0	0	1402	11	52	1.8	67,482	
1	1	3						0	0	0		35685	13.3	47.3 35.5	2.6 3.4	63,667	
0		0				0			0	0	0	42216	15.7	35.5	5	82,455	
0		0				1			0	0	0	7953	13.7	36.6	3.9	64,719	
0		0				0			0	0		29210	16.2	30.5	3.9	64,719	
0		0							0	0	0	29210	16.2	30.9	3.1	60,761	
0		0			0	1	1		0	0	0	14221	12.6	29.9	2.6	63,191	
0		0							0	0	0	4665 2260	11.5	22.4	3.5	52,159	
0		0							0	0	0	1790	13.6	22.4	3.9	51,562 51.562	
0		0				1		1	0	0	0	1650	12.1	38.7 27.2	3.9		
0		0							0	0	0	4355	12.1	17.1	1.7	83,111 62,412	
0		0							0	0	0	4537	7.7	41.4	4.5	62,412 53,590	
0		0							0	0	0	3411	11.5	36.1	3.2	54,735	
0		0	62	.28	0	1	1		0	0		15909	22.9	35.4	2.9	93,833	
0		0		.26	0	1	1	1	0	0		50709	9.8	39	2.2	54,735	
0		0				1	1		0	0		14456	11	51.1	2.9 3.5	86,078	
0		0			0	1			0	0	0	6492	20.2 10.9	49.9	2.9	81,878	
1	4	4			0	1			0	0		37327	12.4	30.6 59	3.2	84,548	
0		0			0	1			0	0		18381	15.1	34.2	4.1	80,702	
0		0			0	1	4		0	0	0	9075	12.7	33.2	2.9	118,494	
0		0			0	1	0		0	0		20126	7.7	58.1	4.3	61,452	
0		0							0	0	0	25467 29973	18.2	22.8	1.8	66,952	
1		3	70						0	0		29973 72512	14.1	28.1	3.3	80,941	
0	1	0				0			0	0	0	7793	14.6	32.5	5	57,443 69,538	
0		0			0		1		0	0	0	2799	17.9	30.8	2.9		
0		0							0	0		11858	10.7	53.3 44.3	3.2	63,141 69,171	
0		0							0	0	0	1729	13.3	44.3	2.4		
1		4							0	0		70641	8.9	33.5	2.6	89,418 61,736	
0	0	0				0	0		0	0		34161	10.1	33.5	2.5	63,981	
0		0				0			0	0		41145	14.1	46.9	2.5 3.4	62,509	
1		2	59	.82		0	0	D	0	0	0	19161	14.4	46.9	3.4	62,509	
1		3				0	0		0	0		19161	12	21.3	3.4	52,278	
0	0	0	61	.17			0	D	0	0	0	15528	12	25	3	79,609	
0		0							0	0		15528	9.7 9.7		3.7	55,635	
		0							0	0	0	2880	9.7		3.7	64,557	
0		0		.93	0	0	0	0	0	0	0	2716					

campmen	tForce used#	ProtestorsBiden	County VoteBlue CountyE	Blue StateBlue	GovernorBlue	County BlueStateBlue	County RedStateRed	County BlueStateRed	County RedStateUni	PopulationPov	erty RateEduc	ationUn	employment	Median Household In	comePrivate Universi	
C)	0	26.78	0 0		0	D	0	0	0	7139	8.1	34.6	1.8	67,809	C
0		0	54.37	0 0			0	0	0			11.5 11.5	42.2 42.2	2.6	74,424	0
0		0	54.37 52.34	0 0				0	0			11.5 10.8	42.2	2.6 2.1	74,424 68,358	1
c		0	53.66	0 0		0	D	0	0		5573	12.9	27.3	5.4	70,838	C
0		0	50.82	0 0			D	0	0	0 2	5000	9.9	33.2	4.1	80,245	0
1		0	54.9 56.53	0 1	(1	1	0	0		3505 3021	8.1 9.3	34.6 38.5	2.6	81,315 82,845	1
Ċ		0	58.05	0 1		1 ·	1	0	0			12.5	41.1	3.9	84,615	1
1	1	3	61.01	0 1	1	1 .	1	0	0		9086	13.5	37.4	3.5	65,034	C
1	0	3	59.74	0 0				0	0			12.9	44.6	2.7	67,906	C
1	1	4	74.82	0 0		1 0		0	0			11.7 10.2	61.8 48.6	2.9 3.4	87,780 80,645	0
c		0	60.84	0 0				0	0			14.8	38.1	3.9	63,822	- 0
C)	0	40.31	0 0	1			0	0	0	9277	27.7	14.6	5	39,931	0
C		0	56.16	0 0			D	0	0			14.8	36.9	3.5	62,992	0
0		0	50.17 41.59	0 0			0	0	0			12.1 12.5	44.8 37.2	3.1 1.7	70,995 64,914	0
1		3	51.11	0 0			0	0	0			14.5	33.1	3.1	70,203	- 0
C		0	20.11	0 0				0	0			15.1	31.5	2.7	71,486	0
0		0	45.15	0 0				0	0		5026	6.1	48.4 34.8	3.5	102,711	0
0		0	64.89 49.31	0 0			0	0	0		1714 3657	13.8	34.8	3.8 3.7	70,871 76,285	- 0
0		0	39.56	0 1				0	0		2990	9.1	32	3.3	90,770	
C		0	53.53	0 0			D	0	0		1264	14.2	30.1	3.2	62,020	
0		0	71.5	0 1			1 D	0	0		1204	11.9	45.5 32.6	2.7	75,041 61,028	
0		0	51.98 41.6	0 0				0	0			14.2	35.6	3.9	70,923	
1			60.46	0 1		1 .	1	0	0	0 2	5681	12.9 14.7	33.4	2.8	64,192	- 0
1	1	4	81.21	0 0			0	0	0	0 4	5008	20.3	34.6	4.2	56,385	
0		0	19.16 26.13	0 0			0	0	0		2651	23.4	16 20.3	5.7	41,837 58,112	
0		0	26.13 35.16	0 0			0	0	0			15.9 10.8	20.3 31.8	4.4 3.5	71,152	
		0	30.71	0 0			0	0	0	0		10.8	23	4.1	54,612	- 1
1	0	4	59.43	0 0	1	1 1	D	0	0	0 4	5541	11.6	44.8	3.3	71,973	
0		0	79.21 38.75	0 1			1	0	0			12.8	48.6 28.7	3.6	79,432 59,184	
1		-	38.75 53.76	0 0			1	0	0			12.1 9.9	28.7 30.1	2.9 4.7	92,793	
0	0	0	54.2	0 1		1	1	0	0			13.2	22.9	4.7	78,779	
0		0	82.33	0 1	1	1	1	0	0	0	584	8.7	60.5	3.3	135,960	
0		0	58.57 21.67	0 1			1	0	0			9.1	50.5 17.9	2.5	99,015 51,449	
1		3	21.67 59.25	0 0	1	1	1	0	0			17.4 13.1	41.2	4.3 3.7	68,169	
c	1	0	43.16	0 0			D	0	0	4	1328 2800	13.1	31.1	3.1	65,967	
0		0	43.16	0 0			D	0	0	0	2800	13	31.1	3.1	65,967 84,551	- 1
0		0	63.06 38.05	0 1		1	1	0	0	0	3027	11.3	41.7 36.6	3.8 2.7	79,117	1
0		0	60.21	0 0			1	0	0			8.3 10.1	42.1	3.9	98,365	
1	0		85.26	0 1	-	1	1	0	0			11.9	60.1	3.3	135,366	
0)	0	53.58	0 0			0	0	0	0	5017	16.1	30.1	3.7	61,168 75,623	- 1
0		0	43.85 37.89	0 0			0	0	0			9.7	34.6 31.1	1.8	67,728	
		0	44.38	0 0			0	0	0			14.1 9.5	44.8	2.8	83,687	
C		0	68.4	0 0			D	0	0	0 4		15.9	41.1	3	61,010	- (
0		0	35.6	0 0			0	0	0	0	6908	14	25.7	2.9	58,535 55,263	- 0
1		0	54.11 52.71	0 0			0	0	0			17.5	46 37.2	2.1 2.9	74,091	1
		0	41.46	0 0				0	0	3		13.2 10.4	34.3	3	72,176	
0)	0	49.44	0 0		0	0	0	0	0		11.3	35.8	2.8	66,427	0
1		4	71.03	0 1 0	1		1	0	0			13.7	35.5	5 3.2	82,455 57,997	
0		0	44.31 66.45	0 0			0	0	0	· 1	1400	13.8 6.9	27.1 52.2	2.4	88,571	
		0	43.45	0 0			0	0	0		9663 7101	19	29.5	3	51,344	
C		0	53.11	0 1	1		1	0	0	0	4062	7.2	36.5	4.2	97,076	1
0		0	71.5 55.94	0 1 0			1 0	0	0	0 1	1572	11.9	45.5 33.7	2.7 4.3	75,041 68,748	
0		0	55.94	0 0			1	0	0		4319	16 10.1	33.7 42.1	4.3	98,365	
c)	0	29.52	0 0			D	0	0			13.2	27.7	3.6	67,978	
0		0	64.89	0 0				0	0	0 1	0487	13.8	34.8	3.8 4.8	70,871 43,604	
0		0	16.62 92.15	0 0			0	0	0			26.9	19.8 63.6	4.8	99,897	
0		0	58.2	0 0				0	0			15.2 14.7	31.5	3.7	65,839	
0		0	24.04	0 0				0	0	0	1016	17.1	18.1	4.4	46,660 84,738	
0		0	55.61	0 1	1			0	0	0	8214	12.6	21.6	6.2	84,738	
0		0	71.03 81.21	0 1				0	0			13.7	35.5 34.6	4.2	56,385	
0		0	19.41	0 1	1		1	0	0		2987 1020	20.3 17	15.3	3.5	59,194	
c)	0	57.39	0 0) (D	0	0	0 2	1610	18.2	28.5	4.2	57,971 91,713	
		3	53.04	0 0				0	0	0 4	7336	9.5	38.6	2.7	91,713 86,579	
1		0	75.78 74.95	0 1	1			0	0		7417	7.8 8.8	55.8 55.9	3.4	116,044	
1 C		0					1	0	0		7023 8838	8.8	55.9	3.4	116,044	
1	1	0	74.95	0 1		1		0	0	0	6056	9.9	30.1	4.7	92,793	
1 0 0 1) 0	2	53.76	0 1	1			0	0					2.2	69 924	
1 0 1 0 0) 	2 0 0	53.76 29.79	0 1 0 0			1		0	0		14.3	22.4	3.3 3.6	69,924 79,432	
1 0 1 0 0 0) 	2 0 0	53.76 29.79 79.21	0 1 0 0 0 1	1	1	1	0	0	0	1516	12.8	48.6	3.3 3.6 2.7	79,432 70,564	
1 0 1 0 0) 	2 0 0	53.76 29.79 79.21	0 1 0 0	1	1 · · · ·	1		0 0 0	0 1	1516 2300	12.8 10.8		3.6 2.7 2.7	79,432 70,564 72,398	
) 	2 0 0 0 0 0 0	53.76 29.79 79.21 54.26 45.49 55.75	0 1 0 0 0 1 0 0 0 0 0 0 0 0		1 · · · · · · · · · · · · · · · · · · ·	1 D D D	0 0 0 0	0 0 0		1516 2300 9769	12.8	48.6 35.7 32.4 36.1	3.6 2.7 2.7 2.6	79,432 70,564 72,398 70,075	
		2 0 0 0 0 0 0 0 4	53.76 29.79 79.21 54.26 45.49 55.75 75.46	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 D D D D	0 0 0 0	0 0 0		1516 2300 9769 1808 3654	12.8 10.8 9.7 12.1 10	48.6 35.7 32.4 36.1 54.4	3.6 2.7 2.7 2.6 2.3	79,432 70,564 72,398 70,075 85,189	
1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0)))))) 1 1 1	2 0 0 0 0 0 0 0 4 3	53.76 29.79 79.21 54.26 45.49 55.75 75.46 69.07	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 D D D D D	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0		1516 2300 9769 1808 3654 8132	12.8 10.8 9.7 12.1 10 17.1	48.6 35.7 32.4 36.1 54.4 33.9	3.6 2.7 2.7 2.6	79,432 70,564 72,398 70,075 85,189 58,375 69,887	
1 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		2 0 0 0 0 0 0 4 3	53.76 29.79 79.21 54.26 45.49 55.75 75.46	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 D D D D D D D	0 0 0 0	0 0 0		1516 2300 9769 1808 3654 8132 5731	12.8 9.7 12.1 10 17.1 10.3	48.6 35.7 32.4 36.1 54.4	3.6 2.7 2.6 2.3 3.8 2.7 3.4	79,432 70,564 72,398 70,075 85,189 58,375 69,887 75,067	
) , 0))) 1 1 1 1 0 0 0 0 0 0 0 0 0 0 0 0 0	2 0 0 0 0 0 0 0 4 3 0	53.76 29.79 79.21 54.26 45.49 55.75 75.46 69.07 46.86 47.55 42.95	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 D D D D D D D	0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0		1516 2300 9769 1808 3654 8132 6731 4999	12.8 10.8 9.7 12.1 10 17.1	48.6 35.7 32.4 36.1 54.4 33.9 31 30.8 24.6	3.6 2.7 2.7 2.6 2.3 3.8 2.7 3.4 2.9	79,432 70,564 72,398 70,075 85,189 58,375 69,887 75,067 63,247	
) , , , , , , , , , , , , ,	2 0 0 0 0 0 4 3 0 0 0 0 0	53.76 29.79 79.21 54.26 45.49 55.75 75.46 69.07 46.66 47.55 42.95 42.95	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 1 0 1 0 6 0 2 0 1 0 1 0 0 0 1 0 0	1516 2300 9769 1808 3654 8132 6731 4999	12.8 9.7 12.1 10 17.1 10.3 11.3	48.6 35.7 32.4 36.1 54.4 33.9 31 30.8 24.6 32	3.6 2.7 2.7 2.6 2.3 3.8 2.7 3.4 2.9 3.5	79,432 70,564 72,398 55,189 58,375 69,887 75,067 63,247 85,508	
		2 0 0 0 0 0 0 4 3 0 0 0 0 0 0 0	53.76 29.79 79.21 54.26 45.49 55.75 75.46 69.07 46.86 47.55 42.25 42.25 42.25 42.01 50.31	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 1 0 1 0 6 0 2 0 1 0 1 0 0 0 1 0 0 0 0	1516 2300 9769 1808 3654 3132 35731 3999 3557 5602 3429	12.8 9.7 12.1 10 17.1 10.3 11.3 13.7 8.3 9.8	48.6 35.7 32.4 36.1 54.4 33.9 31 30.8 24.6 32 33.7	3.6 2.7 2.7 2.6 2.3 3.8 2.7 3.4 2.9	79,432 70,564 72,398 70,075 85,189 58,375 69,887 75,067 63,247 85,508 71,360	
		2 0 0 0 0 0 0 4 3 0 0 0 0 0 0 0 0 0 0 0 0	53.76 29.79 79.21 54.26 55.75 75.46 69.07 46.86 47.55 42.25 42.25 42.25 42.01 50.31 42.07	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 1 0 0 1 0 6 0 2 0 1 0 0 0 0 0 0 0 0	51516 2300 9769 1808 3654 3132 5731 4999 3557 3557 36602 3429 3046	12.8 9.7 12.1 10 17.1 10.3 11.3 13.7 8.3 9.8 10.9	48.6 35.7 32.4 36.1 54.4 33.9 31 30.8 24.6 32	3.6 2.7 2.6 2.3 3.8 2.7 3.4 2.9 3.5 3 3.4 4.3	79,432 70,564 72,398 70,075 85,189 58,375 69,887 75,067 63,247 85,508 71,360 71,434 66,180	
		2 0 0 0 0 0 0 4 3 0 0 0 0 0 0 0	53.76 29.79 79.21 54.26 45.49 55.75 75.46 69.07 46.86 47.55 42.25 42.25 42.25 42.01 50.31	0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0			1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0	0 1 0 1 0 6 0 2 0 1 0 6 0 2 0 1 0 0 0 0 0 0 0 0	515 1516 2300 9769 1808 3654 3655 3656 3656 3657 3656 3657 3656 36577 3657 3657 3657 3657 3657 3657 3657 3657	12.8 9.7 12.1 10 17.1 10.3 11.3 13.7 8.3 9.8	48.6 35.7 32.4 36.1 54.4 33.9 31 30.8 24.6 32 33.7 29.2	3.6 2.7 2.7 2.6 2.3 3.8 2.7 3.4 2.9 3.5 3.5 3 3.4	79,432 70,564 72,398 70,075 85,189 58,375 69,887 75,067 63,247 85,508 71,360 71,434	

0	0	37.65	0	0	0	0	0	0	0	4275	13.9	18.2	3.2	56,599
0	0	58.88	0	1	1	1	0	0	0	3269	13.6	38.1	3.5	70,939
0	0	62.41	0	0	1	0	0	0	0	1928	6.9	52.3	2.8	106,743
0	0	66.36	0	0	0	0	0	0	0	1928	16	35.9	2.8	60,808
0	0	49.24	0	0	1	0	0	0	0	2486	12.7	32.2		
0	0	28.53	0	1	0	0	0	0	0	2486	10.7	40.4	3.4	88,532 76,631
0	0	26.3	0	0	0	0	0	0			9	36.9		
0				0	1		-	0	0	44797	15.7	27.8	2.6	95,085
0	0	41.15	0	1	1	1	0	0	0	5353	15.7	27.8	3.7	63,894
0	0	43.38	0	0	0	0	0	0	0	13567			3.5	52,623
0	0	32.7	0	0	0	0	0	0	0	1930	11.5	30	2.1	67,058
0	0	45.81	0	0	0	0	0	0	0	3919	9.3	30.1	3.7	84,352
0	0	45.81	0	0	0	0	0	0	0	3919	9.3	30.1	3.7	84,352
0	0	53.48	0	1	1	1	0	0	0	2717	9.2	43.4	3.6	106,047
1 1	3	53.89	0	1	1	1	0	0	0	3598	8	39.7	3.3	90,740
0	0	72.03	0	1	1	1	0	0	0	1801	13.8	35.3	4.6	80,180
0	0	60.18	0	1	0	1	0	0	0	1901	9.7	36.5	2	73,057
0	0	62.75	0	0	1	0	0	0	0	13968	10.8	41.6	3.2	83,856
0	0	55.75	0	0	1	0	0	0	0	3012	12.1	36.1	2.6	70,075
0	0	24.5	0	0	0	0	0	0	0	1150	10.9	19.5	2.8	66,283
0	0	41.98	0	1	1	1	0	0	0	2814	13.1	35.8	4.9	91,582
1 1	3	56.16	0	0	1	0	0	0	0	16188	14.8	36.9	3.5	62,992
0	0	70.46	0	1	1	1	0	0	0	54019	10.1	53.3	2.6	89,418
0	0	35.76	0	0	0	0	0	0	0	3384	11.4	25.1	3.8	66,776
0	0	43.79	0	1	1	1	0	0	0	2087	14.9	30.1	4.2	66,542
0	0	56.24	0	0	1	0	0	0	0	1985	8.3	50.2	2.9	90,801
0	0	39.93	0	0	0	0	0	0	0	3121	12.3	25	3.7	64,557
0	0	42.16	0	0	0	0	0	0	0	1205	13	22.6	3.7	61,941
0	0	40.96	0	0	0	0	0	0	0	1966	7.8	35.2	2.5	83,797
0	0	49.95	0	0	1	0	0	0	0	6833	12.6	32.3	2.7	61,255
0	0	37.97	0	0	1	0	0	0	0	1553	9.5	32.8	3.4	70,751
0	0	78.61	0	1	1	1	0	0	0	1231	7.2	60.3	1.9	118,020
1 0	4	57.3	0	1	1	1	0	0	0	4068	7.1	45.5	3.2	84,551
1 0	3	58.22	0	0	1	0	0	0	0	22608	13	40.8	3.7	69,689
1 0	2	60.35	0	1	1	1	0	0	0	17351	12.9	38.1	4.3	78,796
	2	43.79	0	1	1		0	0	0	1873	14.9	30.1	4.2	66,542
0	2	48.86	0				0	0	0	2511	14.1	26	3.7	70,861
1 0	2	72.44	0				0	0	0	3295	14.1	38.3	3.6	71,102
0	3	58.05	U	1	1	1	U	0	U	28878	12.5	41.1	3.9	84,615

References

Allen, Jonathan. 2025. "US immigration agents arrest Palestinian student protestor at Columbia University in Trump crackdown."

https://www.reuters.com/world/us/us-authorities-arrest-palestinian-student-protester-colu mbia-university-students-2025-03-09/.

Arin, Asher and Michael Shamir. "The Primary Political Functions of the Left-Right

Continuum." Comparative Politics 15, no. 2 (1983): 139-58.

https://doi.org/10.2307/421673.

Habeshian, Sareen. 2024. "Exclusive poll: Most college students shrug at nationwide protests." Axios. March 24, 2025.

https://www.axios.com/2024/05/07/poll-students-israel-hamas-protests?utm_medium=soc ial&utm_source=twitter&utm_campaign=editorial

- Chenoweth, Erica, Barton H. Hamilton, Hedwig Lee, Nicholas W. Papageorge, Stephen P. Roll, and Matthew V. Zahn. "Who protests, what do they protest, and why?" *National Bureau of Economic Research*, no. w29987 (2022). <u>https://www.nber.org/papers/w29987</u>
- Chenoweth, Erica, Hammam, Soha, Pressman, Jeremy, and Jay Ulfelder. "Protests in the United States on Palestine and Israel, 2023-2024." *Social Movement Studies* (2024): 1-14. <u>https://www.tandfonline.com/doi/full/10.1080/14742837.2024.2415674?casa_token=nH8</u> <u>C67OX-BYAAAAA%3AuVQHy63o8BUCTAHR7oSXOIMMvbOHRIbpqyPHJim-JlEfJ</u> <u>ws4XEEgwTzCpr1PYFdII0J0Ad_eAV7G</u>.
- "Colleges and Universities." 2023. Opendatasoft. "US Colleges and Universities." Accessed March 24, 2025.

https://public.opendatasoft.com/explore/dataset/us-colleges-and-universities/table/?flg=en_-us.

Dahlum, Sirianne, and Tore Wig. "Chaos on campus: Universities and mass political protest." *Comparative Political Studies* 54, no. 1 (2021): 3-32.

https://journals.sagepub.com/doi/full/10.1177/0010414020919902.

- Datar, Saurabh, Lemonides, Alex, Marcus, Ilana, Murray, Eli, Singer, Ethan, and Christine Zhang. 2025. "An Extremely Detailed Map of the 2024 Election." The New York Times. <u>https://www.nytimes.com/interactive/2025/us/elections/2024-election-map-precinct-result</u> <u>s.html</u>.
- Edwards, Gemma. 2014. *Social movements and protest*. Cambridge University Press. <u>https://books.google.com/books?hl=en&lr=&id=ockNAwAAQBAJ&oi=fnd&pg=PR11&</u> <u>dq=Social+movements+and+protest+Gemma+Edwards&ots=_2-VWo5Esg&sig=r7PqN1</u> <u>IzmigweLuRXtOUD5vI3BI#v=onepage&q=Social%20movements%20and%20protest%</u> <u>20Gemma%20Edwards&f=false</u>
- Eisinger, Peter K. "The conditions of protest behavior in American cities." *American political science review* 67, no. 1 (1973): 11-28. <u>https://doi.org/10.2307/1958525</u>.
- Gallup. 2023. "Democrats' Sympathies in Middle East Shift to Palestinians." (March 1, 2025). https://news.gallup.com/poll/472070/democrats-sympathies-middle-east-shift-palestinian <u>s.aspx</u>.
- Huckfeldt, Robert, Paul Allen Beck, Russell J. Dalton, and Jeffrey Levine. "Political Environments, Cohesive Social Groups, and the Communication of Public Opinion." *American Journal of Political Science* 39, no. 4 (1995): 1025–54.
 https://doi.org/10.2307/2111668.

- Kitschelt, Herbert P. "Political opportunity structures and political protest: Anti-nuclear movements in four democracies." *British journal of political science* 16, no. 1 (1986): 57-85. <u>https://doi.org/10.1017/S000712340000380X</u>.
- Kostelka, Filip, and Jan Rovny. "It's not the left: Ideology and protest participation in old and new democracies." *Comparative Political Studies* 52, no. 11 (2019): 1677-1712. <u>https://hal.science/hal-02386489/</u>.
- McCarthy, John D., Andrew Martin, and Clark McPhail. "Policing disorderly campus protests and convivial gatherings: The interaction of threat, social organization, and First Amendment guarantees." *Social problems* 54, no. 3 (2007): 274-296.

https://doi.org/10.1525/sp.2007.54.3.274.

Meyer, David S. "Protest and political opportunities." *Annu. Rev. Sociol.* 30, no. 1 (2004): 125-145.

https://www.annualreviews.org/content/journals/10.1146/annurev.soc.30.012703.110545.

- Mueller, Carol McClurg. 1992. Frontiers in social movement theory. Yale University Press. https://books.google.com/books?hl=en&lr=&id=2kxcGwv2_u4C&oi=fnd&pg=PR9&dq= Frontiers+in+social+movement+theory&ots=x_oNCrr91J&sig=3aEHpcow6rdaQUeDTs2 F1gbF3Mo#v=onepage&q=Frontiers%20in%20social%20movement%20theory&f=false.
- "North American Industry Classification System." U.S. Census Bureau.

https://www.census.gov/naics/?input=611310&year=2022&details=611310.

PEW Research Center. 2018. "Republicans and Democrats Grow Even Further Apart in Views of Israel, Palestinians." (March 24, 2025). <u>https://www.pewresearch.org/politics/2018/01/23/republicans-and-democrats-grow-even-</u>

further-apart-in-views-of-israel-palestinians/.

- Pierson, P., and Schickler, E. (2020). Madison's constitution under stress: A developmental analysis of political polarization. *Annual Review of Political Science*, 23(1), 37-58. <u>https://www.annualreviews.org/content/journals/10.1146/annurev-polisci-050718-033629</u>
- Popli, Nik. 2024. "Pro-Palestinian Campus Protests Highlight Divisions Among Democrats." Time. <u>https://time.com/6973573/palestine-campus-protests-joe-biden-democrats/</u>.
- Rynhold, Jonathan. "Democrats' attitudes toward the Israeli-Palestinian conflict." *Middle East Policy* 27, no. 4 (2020): 48-61. <u>https://doi.org/10.111/mepo.12526</u>.
- Sabine C. Carey. "The Dynamic Relationship between Protest and Repression." *Political Research Quarterly* 59, no. 1 (2006): 1–11. <u>http://www.jstor.org/stable/4148070</u>.
- Sanders, Austin. 2025. "County-level data sets." U.S. Department of Agriculture, Economic Research Service. (March 24, 2025).

https://www.ers.usda.gov/data-products/county-level-data-sets.

- Schoenmueller, V., Netzer, O., & Stahl, F. (2023). Frontiers: Polarized america: From political polarization to preference polarization. *Marketing Science*, 42(1), 48-60. <u>https://doi.org/10.1287.mksc.2022.1408</u>.
- Silver, J. R., & Shi, L. (2023). Punishing Protesters on the "Other Side": Partisan Bias in Public Support for Repressive and Punitive Responses to Protest Violence. Socius, 9. https://doi.org/10.1177/23780231231182908.
- Tarrow, Sidney. "Progress outside of paradise: Old and new comparative approaches to contentious politics." *Comparative Political Studies* 54, no. 10 (2021): 1885-1901. <u>https://doi.org/10.1177/00104140211024927</u>.
- Tarrow, Sidney. 1998. *Power in movement*. Cambridge university press. https://books.google.com/books?hl=en&lr=&id=OUt6EAAAQBAJ&oi=fnd&pg=PR13&

<u>dq=Power+in+movement&ots=jGftCZNInZ&sig=SER6DF6sQWfWgYo_0woOGwks2u</u> E#v=onepage&q=Power%20in%20movement&f=false.

Thaler, Kai M. "Mixed methods research in the study of political and social violence and conflict." *Journal of mixed methods research* 11, no. 1 (2017): 59-76.

https://doi.org/10.1177/1558689815585196.

- Torcal, Mariano, Toni Rodon, and María José Hierro. "Word on the street: The persistence of leftist-dominated protest in Europe." *West European Politics* 39, no. 2 (2016): 326-350. <u>https://doi.org/10.1080/01402382.2015.1068525</u>.
- Ulfelder, Jay. (2025). "Crowd Counting Consortium U.S. Protest Event Data, 2021-2024." *Harvard Dataverse*. <u>https://doi.org/10.7910/DVN/9MMYDI</u>.
- Vráblíková, Kateřina. "Protest and social movements in political science." *Handbook of social movements across disciplines* (2017): 33-55.

https://doi.org/10.1007/978-3-319-57648-0_3.

Walder, Andrew G. "Political sociology and social movements." *Annual review of sociology* 35, no. 1 (2009): 393-412.

https://www.annualreviews.org/content/journals/10.1146/annurev-soc-070308-120035?ut m_campaign=shareaholic&utm_medium=copy_link&utm_source=bookmark#.

- Williams, Dana M. "How do political opportunities impact protest potential? A multilevel cross-national assessment." *International Journal of Comparative Sociology* 64, no. 4 (2023): 350-374. <u>https://doi.org/10.1177/00207152221133059</u>.
- Zacher, Sam. "Polarization of the Rich: The New Democratic Allegiance of Affluent Americans and the Politics of Redistribution." *Perspectives on Politics*. 22, no. 2 (2024): 338-356. <u>https://doi.org/10.1017/S1537592722003310</u>.