

Who Supports Political Violence?

Miles T. Armaly

Assistant Professor

University of Mississippi

Department of Political Science

133 Deupree Hall

University, MS 38677

mtarmaly@olemiss.edu

Adam M. Enders

Assistant Professor

University of Louisville

Department of Political Science

Ford Hall, Room 205

Louisville, KY 40292

amende01@louisville.edu

Abstract

The last few years have witnessed an increase in democratic “backsliding” in the U.S.—a decline in the quality of democracy, typically accompanied by an influx of non-normative behavior, such as political violence. Despite the real consequences of support for violence, fairly little is known about such an extremist attitude outside studies of terrorism or aggression. Using a unique survey containing many psychological, political, and social characteristics, we find that perceived victimhood, authoritarianism, populism, and white identity are the most powerful predictors of support for violence, though military service, conspiratorial thinking, anxiety, and feelings of powerlessness are also related. These patterns suggest that subjective feelings about being unjustly victimized—irrespective of the truth of the matter—and the psychological baggage that accompanies such feelings lie at the heart of support for violence. We use these results to build a profile of characteristics that explain support for violence; the predictive validity of this profile is then tested by examining its relationship with support for the January 6, 2021 U.S. Capitol riot, which it is strongly associated with, even accounting for support for Donald Trump. Our findings have implications for the detection of extremist attitudes and our understanding of the non-partisan/ideological foundations of anti-social political behavior.

Keywords: violence, populism, conspiracy theory, victimhood, authoritarianism

Word count: 10,228

A January 2021 American Enterprise Institute survey of U.S. adults found that more than a third of Americans agree that “the traditional American way of life is disappearing so fast that we may have to use force to save it,” including 56 percent of Republicans, 22 percent of Democrats, and 35 percent of independents.¹ More than an abstraction, hostility and support for political violence have measurable behavioral consequences—this was never clearer than on January 6, 2021, when hundreds of individuals forced their way into the U.S. Capitol Building, disrupting the certification of a presidential election. While few may be willing to participate in physical violence, that so many are willing to entertain or support the use of force in order to preserve the “American way of life” is disconcerting.

Although some is known about the macro-political conditions that foster political violence, as well as the social and demographic characteristics of violent extremists, less attention has been paid to extremist attitudes among the mass public. In a sense, this is for good reason—less polarized times required less attention to such a topic. Times have changed. Growing anger, at both out-groups and the government more generally, has resulted in record low levels of political trust (Webster 2020) and a sharp increase in hostility toward out-groups (Iyengar and Krupenkin 2018). More Americans supported the use of tear gas on out-group protestors in 2017 than in 2014 (Westwood, Peterson and Lelkes 2019). A majority of Americans recognized the potential impact of heated political rhetoric on the prospect of political violence in 2019, before the pandemic, the impeachments of Donald Trump, or the Capitol riots.² Lowering the temperature of polarization, combating uncivil attitudes, and preventing political violence all require a better understanding of the characteristics and motivations behind hostile, non-normative political attitudes and behavior.

In this paper, we examine the predictors of support for political violence in order to build a predictive profile of individuals who are likely to support violence (i.e., people who agree with statements like, “Violence is sometimes an acceptable way for Americans to

¹<https://www.americansurveycenter.org/research/after-the-ballots-are-counted-conspiracies-political-violence-and-american-exceptionalism/>

²<https://www.pewresearch.org/fact-tank/2019/07/18/americans-say-the-nations-political-debate-has-grown-more-toxic-and-heated-rhetoric-could-lead-to-violence/>

express their disagreement with the government”). The 24 correlates we employ span the psychological, political, and social domains. We find that feelings of victimhood, authoritarianism, populist sentiments, white identity, and military service are the strongest correlates of support for violence, though anxiety, feelings of powerlessness, conspiratorial thinking, perceived governmental corruption, racial resentment, and religiosity are also related. These results demonstrate that psychological orientations and attitudinal postures are more useful in explaining support for violence than sociodemographic-based constructs, such as socioeconomic deprivation or status threat, as is sometimes assumed of those who participate in political violence. Instead, more subjective feelings that one is being unjustly victimized, as well as postures that might help resolve such injustices—like support for authoritarian and populist sentiments—are more predictive. In other words, one need not experience tangible oppression or subjugation to be supportive of extremist, violent action against perceived oppressors.

Finally, we use these results to build a profile of individuals who are most (least) likely to support political violence. To test the predictive validity of the profile, we use it to explain attitudes about the U.S. Capitol riot (e.g., “The riots at the U.S. Capitol building on January 6th were justified”). The political violence profile is effectively a summary of the foundational attitudinal and psychological antecedents of support for violence. Employing the profile allows us to avoid the circular logic of explaining support for a specific violent event with support for abstract violence; rather, we are accounting for the complex constellation of correlates that underlies support for violence. We find that the political violence profile exhibits a stronger relationship with support for the riot than any other explanatory variable, although positive feelings about Donald Trump are also an important factor. Moreover, we observe a statistically significant interaction between Trump support and the political violence profile: those who are highest on the political violence profile exhibit the strongest connection between Trump support and support for the Capitol riot. Thus, approval of real world political violence appears to be the product of a toxic blend of partisan attachments

and several non-political orientations that encourage violence.

Our results are important for several reasons. First and foremost, the specter of the January 6 attacks on the U.S. Capitol looms large. So, too, do threats like the kidnapping plot against Michigan Governor Gretchen Whitmer.³ Although few participated in the riots or kidnapping plot, relatively speaking, mass political violence may now be less abstract in the minds of ordinary Americans. Developing a robust understanding of who supports such actions—both abstractly and concretely—is useful for social scientists and law enforcement alike. Yet, our focus is not exclusively on the January 6 attacks, which were highly partisan in nature and inflamed by elected leaders. The Capitol riots were merely the starkest, most recent display of political violence that occurs in the U.S., albeit on smaller stages. For instance, the 2016 election season was marked by many smaller acts of violence.⁴ Moreover, politically motivated violence is on the rise nationwide.⁵ As the AEI survey referenced above notes, many Americans may now see violence as a legitimate form of political expression and participation, even if particular individuals would be unlikely to participate in violent acts. Thus, it is especially useful for scholars of mass political behavior to understand the constellation of factors that underwrites support for violence, particularly in a contemporary political environment characterized by extreme rhetoric and violence.

Known Correlates of Violence

Several literatures spanning academic disciplines and levels of analysis provide hints at the characteristics of individuals who are likely to support political violence, even if little work has directly examined this question. At the macro-political level, there is a correlation between economic inequality and actual political violence across political and cultural contexts (Sigelman and Simpson 1977). This relationship also manifests at the individual level: socioeconomic deprivation—operationalized as low levels of education and

³<https://thehill.com/homenews/state-watch/520189-fbi-says-it-foiled-plot-to-kidnap-michigan-governor>

⁴<https://www.wsj.com/articles/election-2016-a-violent-season-marked-with-arrests-1478552509>

⁵<https://www.washingtonpost.com/investigations/interactive/2021/domestic-terrorism-data/>

income—is related to support for violence (Canetti et al. 2010).

Political violence research is overwhelmingly concerned with the impact of religion, especially as many of the instances of terrorism observed across the globe are at least partially motivated by religious principles and movements. At the individual level, however, there is contention about the precise role of religion in either participation in, or support for, political violence. Fair and colleagues (2012), for example, find that only Pakistanis who believe jihad can and must be waged by individuals are supportive of political violence—religious practice and support for political Islam are not, on their own, sufficient for motivating support for violent groups. Similarly, Canetti et al. (2010) find that the effect of religiosity on support for political violence is mediated by socio-economic deprivation. Thus, while there does appear to be some connection between religiosity and support for violence, other conditions must also be met for the relationship to take place.

Social psychological work on the nature and consequences of aggression and anger also provides insight into likely correlates of attitudes about violence. Individuals who exhibit elevated predispositions toward aggression tend to also exhibit frustration with their lives, congruent with other theories about the impact of socioeconomic deprivation on support for violence (Anderson and Bushman 2002). Men are also more aggressive, on average, than women (Anderson and Bushman 2002). More than a “bottom up” psychological process, aggression can be activated and inflamed by violent political rhetoric, thereby increasing support for political violence (Kalmoe 2014). Strong partisans—who, by virtue of interest in and attention to politics, are most likely to be exposed to violent political rhetoric—hold the most uncivil attitudes about political out-groups (Miller and Conover 2015). Those exhibiting the most hostile attitudes and elevated levels of anger about politics are both more likely than others to participate in politics and less committed to democratic norms and values (Webster 2020)—a toxic blend presumably capable of fueling support for political violence.

In addition to political, socioeconomic, and purely psychological factors, a host of

social-psychological orientations regarding power, authority, and trust are related to support for, or engagement in, political violence. Aggression, which underlies violence, is part and parcel of the authoritarian personality (Altemeyer 1981), a willingness to submit to authority at the expense of societal freedom (Adorno et al. 1950). Faragó, Kende and Krekó (2019) demonstrate that authoritarianism more strongly predicts justification for violence against both symbolically threatening and physically dangerous groups than the propensity for radical action. That is, strong support for order and authority weigh more heavily on support for violence than one’s willingness to, for instance, “participate in a violent act to defend your opinion or values.” Thus, there is specific precedent for authoritarian sentiments to relate to support for violence.

In this same vein, populist sentiments, which are similar in many regards to authoritarian ones (Norris 2005), have been connected to political violence. Populism, an “us-versus-them worldview” where “us” is the masses and “them” are the elites (Berman 2021), is partially driven by social and economic grievances and the perceived inattention of political elites to the plight of the masses (Piketty 2017). These grievances are then subject to manipulation, resulting in the direction of anger and violence at a specified target (Berlet 1995). Additionally, support for populist sentiments is related to behavioral violence; for example, those who support the recent “America first” variant of populism are likelier to have been arrested on criminal charges throughout their lives (Levi, Sendroiu and Hagan 2020). Indeed, Snyder (2018) notes that, historically speaking, “violence is hugely important” in the current era of populism.

Finally, conspiratorial thinking has been linked to support for violence in several ways. Uscinski and Parent (2014) find that conspiratorial thinking—the general predisposition to interpret salient events and circumstances as the product of conspiracies—is positively related to the abstract support for violence. Similarly, Imhoff, Dieterle and Lamberty (2021) demonstrate that individuals prone to conspiratorial thinking are less likely to participate in normative political activities, like voting or contacting representatives, and are more likely

to engage in non-normative activities, such as committing a violent attack on a person in power. These connections bear out even in less abstract scenarios: those who believe that 5G wireless technology was spreading Covid-19 were more likely to support violent measures to stop the spread, such as arson attacks on 5G cellular towers (Jolley and Paterson 2020). Finally, beliefs in a number of specific conspiracy theories are correlated with support for political violence (e.g., Enders et al. 2022; Uscinski et al. 2021), and others have observed a connection between positive feelings regarding the QAnon movement and support for the Capitol riot, in particular (Armaly, Buckley and Enders forthcoming).

While disparate literatures across several disciplines have hinted at the psychological and social factors that underlie support for the use of violence, these idiosyncratic research programs typically fail to control for the full complement of psychological and social factors. By and large, the overarching goal of extant literature has been to identify individual or small groups of predictors of attitudes about violence, rather than learning who is most likely to support violence. Even though violence is fairly well-studied, it is unclear which of the identified correlates matters most when it comes to supporting violence. This problem is exacerbated by the fact that many of the identified predictors are likely to overlap with one another. Populism, conspiratorial thinking, and authoritarianism, for example, tend to be highly correlated. Likewise, socioeconomic deprivation is, definitionally, a confluence of personal circumstances regarding income, education, and employment. Finally, people who exhibit higher levels of conspiratorial thinking and populism tend to exhibit the kind of feelings—like helplessness and powerlessness (e.g., Jolley and Douglas 2014)—that socioeconomic instability promotes.

What’s Missing?

Our central goal in this manuscript is twofold. First, we endeavor to build off previous work by expanding the list of potential explanations for support for political violence. Recent

violence in the United States—most notably, the January 6, 2021 U.S. Capitol riots—has taken place in a climate of impassioned *distrust* and perceived *corruption* among a group of mostly *white men* who believed they were the *victims* of electoral fraud and democratic subversion. Thus, we see the potential for many additional predictors of support for violence. Second, we aim to determine which of these factors—many of which overlap and may interact with one another—appear to be most important to explaining support for violence.

The last few years have witnessed a sharp influx in the role of populist, authoritarian, and conspiratorial rhetoric in political communication (e.g., Hameleers and Vliegenthart 2020). The predispositions at the center of these rhetorical strategies are driven by impassioned displeasure with the established political order and other out-groups—sentiments that we suspect may also lead one to be accepting of political violence, if not only for the vast array of violent anti-governmental groups in America.⁶ As such, we investigate the relationships between support for violence and populism, conspiratorial thinking, authoritarianism, perceived governmental corruption, and trust in government. While, as noted above, we are far from the first to consider the role populism, authoritarianism, and conspiratorial thinking play in aggression and violence, we argue it is important to relate the factors, directly, to attitudinal support for violence at the individual-level.

We also operationalize frustration in several ways that are not explicitly anchored to politics or government. Perceptions of victimhood, for example, are born of a blend of entitlement and a feeling that one is not receiving what they deserve (Armaly and Enders 2021). We utilize two forms of perceived victimhood: egocentric and systemic. The egocentric variant captures general feelings that one is always settling for less. The systemic variant of perceived victimhood captures sentiments regarding “the system,” specifically, working against the individual. Armaly and Enders (2021) demonstrate that each form of victimhood is related to support for policies and politicians that can remedy the perceived victimhood or the cause thereof. In a similar fashion, we expect perceived victimhood to positively relate to

⁶<https://www.splcenter.org/fighting-hate/extremist-files/ideology/antigovernment>

support for violence—violence may be seen as a way to remedy one’s victim status, to wrestle control of one’s life back from the oppressive powers that be. Additionally, we account for anxiety and feelings of powerlessness, which can serve as purely psychological markers for a host of frustrations with the political system or particular groups (e.g., Douglas et al. 2019).

We also expect that racial identity and prejudice—central cleavages in American politics—plays a role (Jardina 2019). We account for each separately. White identity refers to the degree to which one feels attached to their whiteness and solidarity with other whites. Jardina (2019) argues that white identity arises, in part, because of status threat as a reaction to growing diversity. Violence may be seen as a way to mitigate potential status loss, or to preserve the existing racial hierarchy from which they perceive themselves to benefit. Therefore, we expect those high in white identity to support political violence. Likewise, racial resentment—our operationalization of racial prejudice—represents both perceived violations of American values of hard work, as well as negative affect toward blacks (Kinder and Sanders 1996). We expect those high in racial resentment to be more supportive of political violence. First, prejudice is clearly linked to conflict and violence (Green and Seher 2003; Green, Strolovitch and Wong 1998), so there is sound reason to suspect it relates to support for violence. Second, much like our expectations regarding white identity, violence may be considered a viable way to stave off the advancement of minorities and preserve status for those perceived to uphold individualism and hard work.

Finally, we consider whether or not one has access to health insurance, as well as the potential impact of military service. We employ access to health insurance as an operationalization of socioeconomic deprivation—which should be positively related to support for violence, based on previous literature (e.g., Anderson and Bushman 2002)—that does not rely on self-reported income, an error-prone measure for which there also tend to be relatively high levels of missing data (e.g., Moore, Stinson and Jr. 2000). We examine military service because both veterans and active military personnel are more likely to engage in partner violence (Kwan et al. 2020) and anti-governmental groups, such as the Oath Keepers

and Three Percenters, tend to be composed disproportionately of individuals with military service records (primarily veterans).⁷

In addition to these newly identified correlates, we reexamine many of the correlates others have identified (e.g., religiosity, income, and education). If our goal is to decipher which (combinations of) correlates best explain support for political violence, our model must incorporate as many of such correlates as possible.

Data & Measures

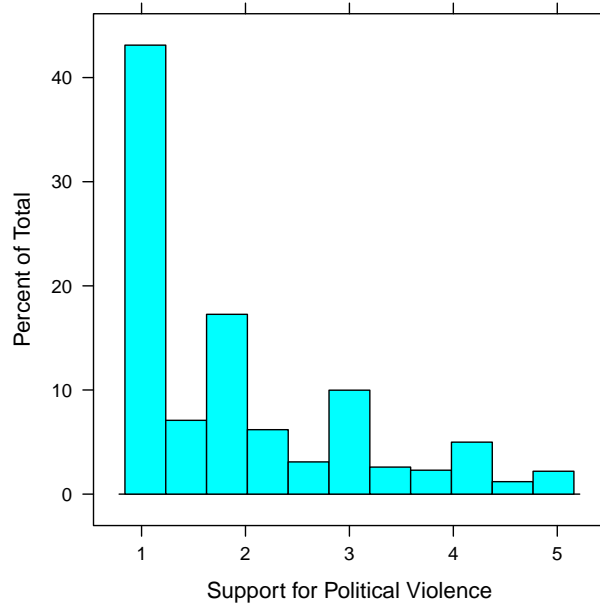
We fielded a survey containing the questions necessary to construct 24 possible correlates of political violence on 1,100 U.S. adults in February 2021. Our sample was fielded by Lucid, which is a survey marketplace that recruits respondents from dozens of sample suppliers. While the sample is not a probability sample, the demographic characteristics of respondents match those of the broader population according to U.S. Census data⁸ and the platform has been found to perform very well in generating accurate estimates of various political attitudes (Coppock and McClellan 2019). In light of recent evidence regarding inattention on survey platforms like Lucid—in particular during the COVID-19 pandemic (Aronow et al. n.d.; Peyton, Huber and Coppock 2020)—we took several steps to ensure data quality beyond Lucid’s use of reCAPTCHA to screen for bots and a combination of open-ended questions and machine learning methods to screen out inattentive respondents. First, respondents had to pass two additional attention checks in order to complete the survey. Second, we restricted our sample to only those respondents that spent an appropriate amount of time—at least 7 minutes, on a survey estimated to take 15 minutes—answering our questions. The sample size after this restriction is 1,002.

Our dependent variable, which was originally developed by Uscinski and Parent

⁷<https://www.splcenter.org/fighting-hate/extremist-files>

⁸See the appendix for a comparison of the sociodemographic characteristics of the sample with the most recent U.S. Census data. We also compare partisan and ideological identities between our sample and the 2020 ANES.

Figure 1: Distribution of support for political violence.



(2014), is a summated scale of responses to the following three questions about support for political violence, each of which respondents reacted to using a five-point set of responses ranging from “strongly disagree” (1) to “strongly agree” (5):

1. It is acceptable to use violence in advancing political goals these days.
2. Violence is sometimes an acceptable way for Americans to express their disagreement with the government.
3. Violence is justified if the members of the other side act violently first.

Importantly, Uscinski and Parent (2014) suggest these items tap general support for and the perceived acceptability of violence as a way of achieving one’s political goals, rather than a willingness to engage in violence against any one group (e.g., party), the government in particular, or for any specific cause. In addition, these items avoid some of the problems identified by Westwood et al. (2021) that inflate support for political violence (e.g., unbalanced options with no midpoint), particularly as caused by inattentive respondents

(Peyton, Huber and Coppock 2020).⁹ The scale ($M=1.91$, $SD=1.10$) is statistically reliable ($\alpha=0.86$). As Figure 1—which depicts the distribution of the violence scale—shows, most individuals are not supportive of the use of violence in politics. The modal scale value is 1, which corresponds with strong disagreement with each of the above three questions; this characterizes 43% of respondents. Still, 13% of respondents agree more than they disagree with the use of political violence and 10% are completely neutral, figures that comport nicely with those reported in other work on the appropriate measurement of support for violence (see Westwood et al. 2021). For example, the mean of our measure (rescaled to range from 0–1 for comparability) is only between 0.08 and 0.03 points higher than what Kalmoe (2014) reports across his three measures, which include more specific imagery about using bullets and bricks to solve political problems. Moreover, the mean of our measure is statistically indistinguishable at conventional levels ($p = 0.07$) from that of a measure of general willingness to engage in violence reported by Jolley and Paterson (2020).¹⁰ Thus, it does not appear that our measure of support for political violence is merely capturing attitudes regarding the Capitol riot that occurred in the month prior to data collection. Given the potential consequences of attitudes about violence, we still see reason to be concerned with the people’s attitudes, even if most Americans do not support the use of political violence.

Summary statistics for each correlate under consideration appear in Table 1; see the appendix for precise question wording. We organized these correlates by category, from general psychological traits and orientations to sociodemographic characteristics and individual circumstances. Each of the multiple-item scales we employ—all of which have been

⁹We note, however, that debate continues about the most appropriate way to measure support for violence, which is itself surely tied to one’s research question. For example, Westwood et al. (2021) show that inferences about how many people support political violence depends heavily on question wording. Ours should not be taken as the definitive word on this subject—we strongly encourage replication of our analyses using different measurement strategies.

¹⁰We also undertook an investigation of potential respondent satisficing, which could inflate our estimates of support for political violence or potentially inflate the correlation between support for political violence and the independent variables we consider. We find evidence for neither concern. Details of these analyses appear in the appendix.

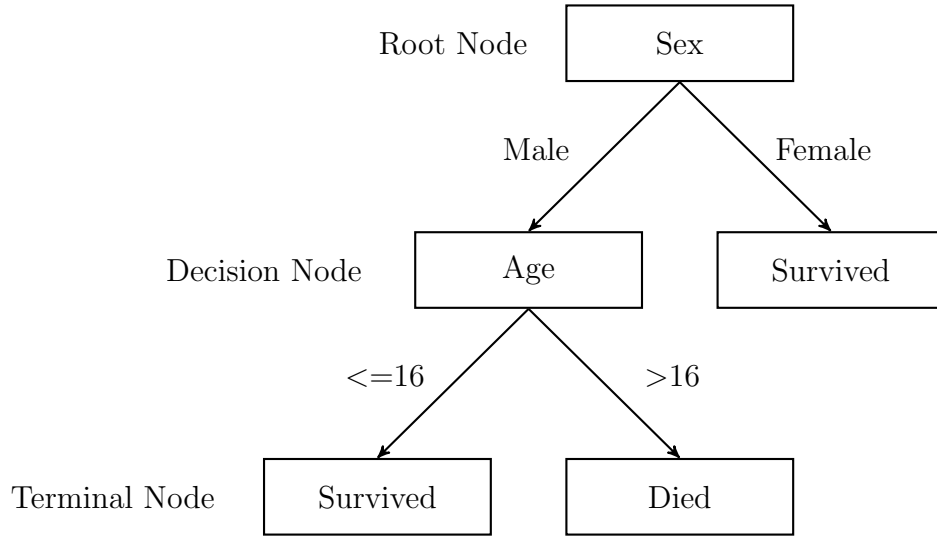
Table 1: Descriptive statistics for all potential correlates of support for political violence.

Category	Variable (Range)	Mean	Std. Dev.	Alpha
General Psychological Traits and Orientations	Perceived Systemic Victimhood (1-5)	2.65	1.04	0.83
	Perceived Egocentric Victimhood (1-5)	2.70	1.08	0.89
	Anxiety (1-4)	1.84	0.83	0.94
	Powerlessness (1-5)	2.30	0.80	0.78
Orientations Toward Government and Politics	Authoritarianism (1-5)	2.91	0.90	0.65
	Populism (1-5)	3.33	0.80	0.83
	Conspiratorial Thinking (1-5)	3.45	0.94	0.84
	Perceived Corruption (1-5)	3.16	0.98	—
	Trust in Government (1-5)	2.47	0.99	—
Racial Attitudes	Interest in Politics (1-4)	2.94	0.95	—
	White Identity (1-5)	3.28	0.95	0.81
	Racial Resentment (1-5)	2.94	1.05	0.79
(Strength of) Political Identities	Ideological Strength (1-4)	2.45	1.15	—
	Partisan Strength (1-4)	3.00	1.11	—
	Republican (0, 1)	0.35	0.48	—
	Conservative (0, 1)	0.37	0.48	—
Religious Characteristics	Religiosity (1-5)	2.50	1.57	—
	Evangelical (0, 1)	0.26	0.44	—
Sociodemographics and Individual Circumstances	Military Service (0, 1)	0.15	0.35	—
	Income (1-5)	3.35	1.24	—
	Health Insurance (0, 1)	0.88	0.32	—
	Education (1-5)	3.30	1.06	—
	Female (0, 1)	0.53	0.50	—
	Residence in South (0, 1)	0.37	0.48	—

previously validated by others (see appendix)—are statistically reliable.

Our analytical strategy unfolds in three steps. First, we examine bivariate correlations between support for violence and each of the potential correlates in Table 1. Second, we use a classification and regression tree (CART) model to decipher which variables help us best predict support for violence. The goal of CART models is to correctly classify values of a specified outcome variable (support for violence), allowing for complex nonlinear and interactive relationships between the predictor variables. To understand how CART models work, one might imagine a simple decision tree. We present a hypothetical example in Figure 2. Suppose we were attempting to model whether or not one survived the sinking of the Titanic, a classical example used to explain how CART models work. Because of the

Figure 2: Hypothetical example of a CART model.



“women and children first” evacuation policy on the Titanic, we know that sex and age are two of the most important variables in classifying whether or not one survived. Indeed, sex—represented by the “root node” at the very top of the tree—is the most predictive factor. If one was female, they were very likely to survive, which is represented by the rightmost path leading from the sex node to the survived terminal node. Terminal nodes represent the final prediction for a branch of decisions. If one was male, we must further wonder they were a child or not. Males less than or equal to 16 years of age were likely to survive, as depicted by the series of leftmost paths from sex to the age decision node, and the age decision node to the survived terminal node. Males greater than 16 years of age were likely to perish, as depicted by the path from sex to the age decision node, and the age decision node to the died terminal node. In the immediate study, we seek to predict one’s level of support for violence, which will be depicted in terminal nodes, using the 24 correlates discussed above, which will be depicted in the root node (simply the single most predictive factor, like sex in the Titanic example) and a series of decision nodes (like age).

The CART modeling procedure—implemented in `party` R package (Hothorn, Hornik and Zeileis 2006)—first identifies a single independent variable that best organizes the data into two groups; “best” is defined as the split in the variable that explains the greatest

variance in the outcome variable. Once the data are separated, this process is recursively applied to each subgroup until no improvements can be made (i.e., until no more variance is explained). The procedure sequentially selects the variables that best improve the fit of the model to the dataset. Thus, it does not require linear relationships between the outcome and predictor variables. Partitioning the data at a particular level of a predictor variable—for example, a 3 (the midpoint) on the anxiety scale—allows for the consideration of the impact of variables at differing levels of other covariates.¹¹

Finally, we examine the predictive power of the profile of orientations that the CART model identifies as uniquely useful in classifying attitudes about violence. In particular, we attempt to explain support for the U.S. Capitol riots—“The riots at the U.S. Capitol building on January 6th were justified”—using a combination of political identities, positive feelings toward Donald Trump, and the political violence profile. Our expectation is that the political violence profile is a strong predictor of these attitudes, even controlling for other relevant factors. Moreover, the profile should be more strongly related than any individual correlate of support for political violence if our strategy has yielded a valid and predictive picture of who supports political violence.

To be clear, this is not merely a more complicated way of controlling for support for violence. Instead, the political violence profile is a summary measure of the psychological, orientational, and attitudinal antecedents of political violence. We argue this approach is useful for a number of reasons. First, it avoids the tautology of predicting support for the Capitol riot with support for violence, generally. Instead, we can state, for instance, “individuals with the following *collection* of orientations are the most likely to state the riot was justified.” Second, this approach offers analytical strength in that there is no assumption of functional form between the 24 correlates detailed below and support for the Capitol riot.¹²

¹¹The procedure does not inherently dichotomize the variable when it partitions the data. See below for more detail.

¹²There is, of course, a functional form assumption between the political violence profile and support for the riot. However, how one gets categorized into the levels of the profile via the CART model is not assumed to be linear, nor is it linear empirically, as we demonstrate below.

It is possible that high levels of, say, powerlessness contribute to both high violence support and low violence support. Accounting for either violence, itself, or powerlessness in a regression model explaining Capitol riot justification would mask such a nuanced relationship. We expect—and, per the results of our CART model, find support for the notion—that the interactive and nonlinear effects of several correlates of support for violence are important pieces of the puzzle we are attempting to solve (i.e., who supports political violence?). Third, this approach constitutes a more parsimonious accounting of support for the Capitol riot than a “garbage can” model of all possible explanations. Rather than estimate nearly 30 parameters, we have empirical reason to condense a number of correlates into a simpler, albeit analytically powerful, profile. This allows for a cleaner, more straightforward understanding of who supports violence, generally, and the Capitol riot, specifically.

Results

Figure 3 displays pairwise correlations between support for violence and each potential correlate. The two forms of perceived victimhood, support for authoritarianism and populism, and white identity are the strongest correlates of support for violence. Those with (seemingly) more extreme general psychological traits and orientations (e.g., powerlessness and anxiety) exhibit the highest levels of support for violence. The same is true of those who have an antagonistic relationship with government—those who think it is corrupt, engage in conspiratorial thinking about nefarious actors, and support non-traditional leaders, like authoritarians or populists. But, a number of other characteristics correlate with support for violence. For instance, evangelicals and those who frequently attend religious services are more supportive of violence. So, too, are current and former military personnel.

Political predispositions, on the other hand, do not correlate with support for violence. Congruent with past literature, there is a correlation between strength of ideological identification (and partisan strength is marginally significant¹³), but it pales in magnitude

¹³One reason why the strength of partisan identification may not be significant here, though it is in other

compared to most of the other correlations described above. Altogether, support for violence does not appear to stem from a single attitude or circumstance; instead, a constellation of factors underwrite it.

However, it is likely that many of these characteristics, attitudes, and orientations covary with one another. Indeed, we see a number of moderate and strong relationships in Figure 4, which displays pairwise correlations between all of our covariates. Consider, mostly notably, the bottom left portion of Figure 4. The psychological and government-specific orientations are, generally speaking, strongly correlated with one another. Inasmuch as our goal is to determine which of these covariates combine to best characterize individuals high in support for violence, correlational analyses may prove deficient.

We also observe considerable nonlinearity between support for violence and each of the independent variables. In the appendix we present scatterplots for each relationship, overlaying both nonparametric smoothers and OLS fit lines. In approximately half of the cases, we observe OLS fit lines that are not comfortably encompassed by the confidence intervals of the nonparametric smoother, which is indicative of a statistically significant deviation from linearity. For instance, as Figure 5 shows, the relationship between racial resentment and support for violence is, according to the nonparametric smoother, sinusoidal rather than linear. That is, those of middling levels of racial resentment are significantly more supportive of political violence than those either high or low in racial resentment. Such a relationship presents unique challenges in relating racial resentment to violence in the multivariate context when a (linear) functional form is assumed. The profile approach offered by CART models allows for us to properly incorporate this nonlinear relationship into our model of support for violence.

To aid in the identification of a profile that predicts support for violence, we turn to a CART model, the results of which are presented in Figure 6.¹⁴ Because these models are not work, is because none of the three items composing the support for violence scale directly mention parties or other partisan groups.

¹⁴Model $n=1,002$.

Figure 3: Pearson product-moment correlation with support for violence, with 95% confidence intervals. $n=817-1,002$

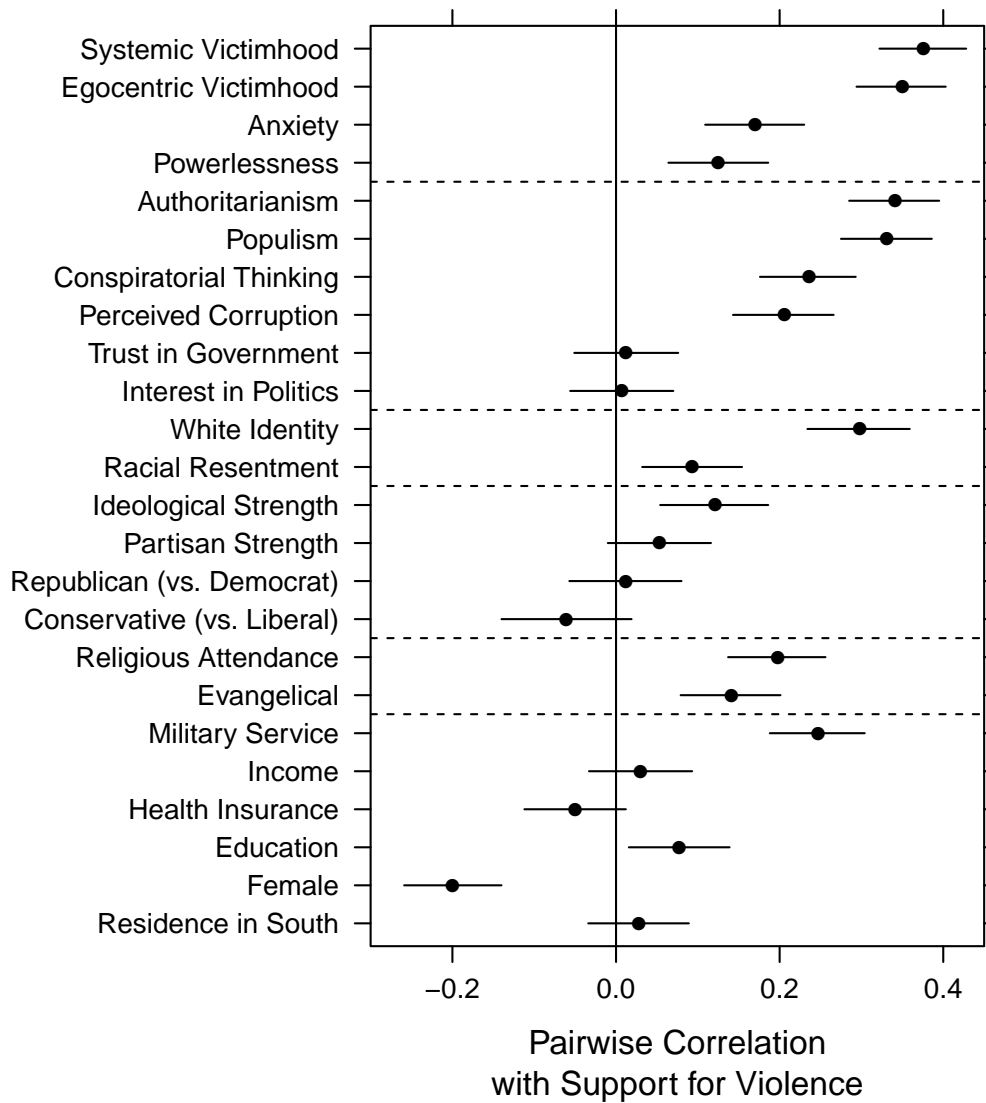


Figure 4: Pairwise correlation heatmap.

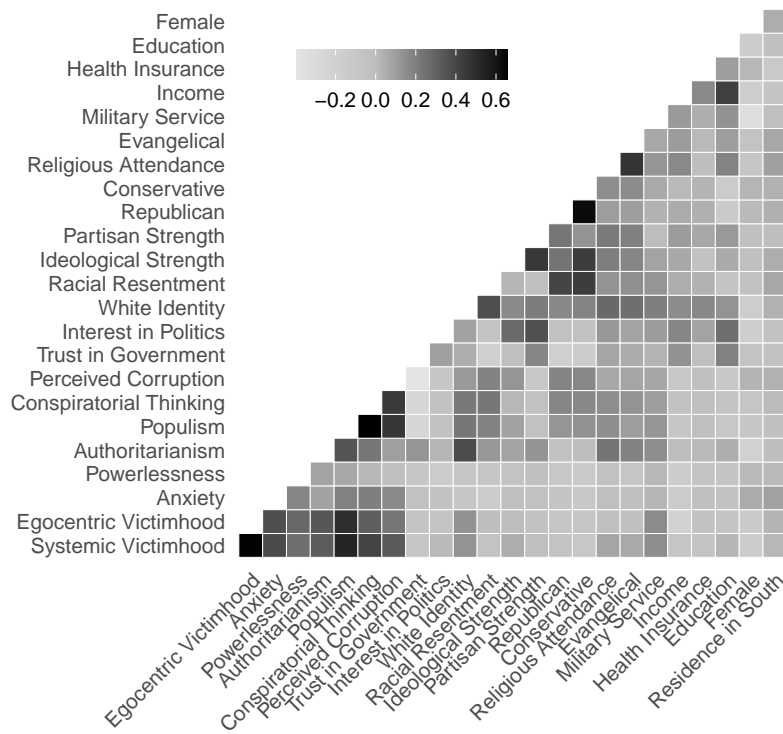
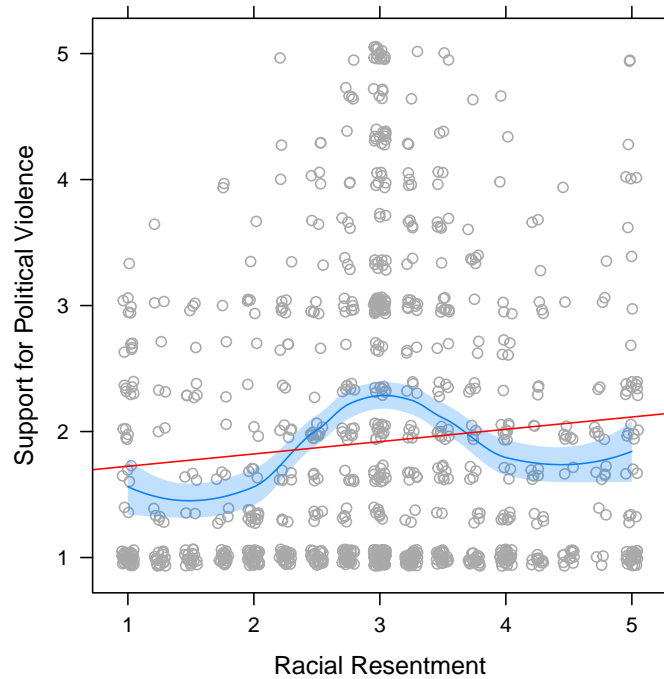


Figure 5: Support for political violence across racial resentment. Red line indicates OLS fit line; blue line nonparametric smoother.



common in political science, we carefully walk the reader through one set of nodes, and note that all remaining paths and nodes can be interpreted in a similar fashion. The distribution of support for political violence for various combinations of independent variables appears at the very bottom of the figure. Each boxplot across the horizontal axis details the level of support for violence for respondents classified into that node. We are interested in both the distribution of violence support in each terminal node, as well as how one came to be classified into any given node.

First, the CART procedure identified the predictor variable that is the most highly predictive of support for violence, which appears at the top of the tree: perceived systemic victimhood. Then, systemic victimhood was partitioned at the value that explains the greatest variance in support for violence; here, that value is 3.5 (which is slightly greater than the midpoint, 3). Importantly, we note that the method is not inherently dichotomizing the variables.¹⁵ Following the left path, we are considering individuals with systemic victimhood scores of 3.5 or less. Then, the remaining data are once again split using the variable that explains the greatest remaining variance: populism (Node 2). Populism is partitioned at 2.625 (which is just lower than the midpoint, 3). Again following the left node, racial resentment explains the greatest remaining variance (Node 3), and is itself partitioned into those with scores greater than 2 (a low-medium score), or less than or equal to 2. Thus, for this collection of 184 individuals in our dataset, three variables—systemic victimhood, populism, and racial resentment—best classify their support for violence (which, per the boxplots at Nodes 4 and 5, is very low). These variables all interact with one another to best classify support for violence. Each of the factors after the first node is dependent upon systemic victimhood, and every other node along the way.

¹⁵Even though it partitions each variable somewhere along the scale, this is 1) empirically determined and 2) variables can be re-partitioned many times as the algorithm moves down the tree, not just once. The implication of the former point is that variance is not being arbitrarily wasted; rather, meaningful variance is being efficiently utilized while less meaningful variance (some of which might even constitute measurement error) is being ignored. The implication of the second point is that the partitioning process is precisely what allows for the incorporation of nonlinear and interactive relationships that we turn to the CART model to investigate.

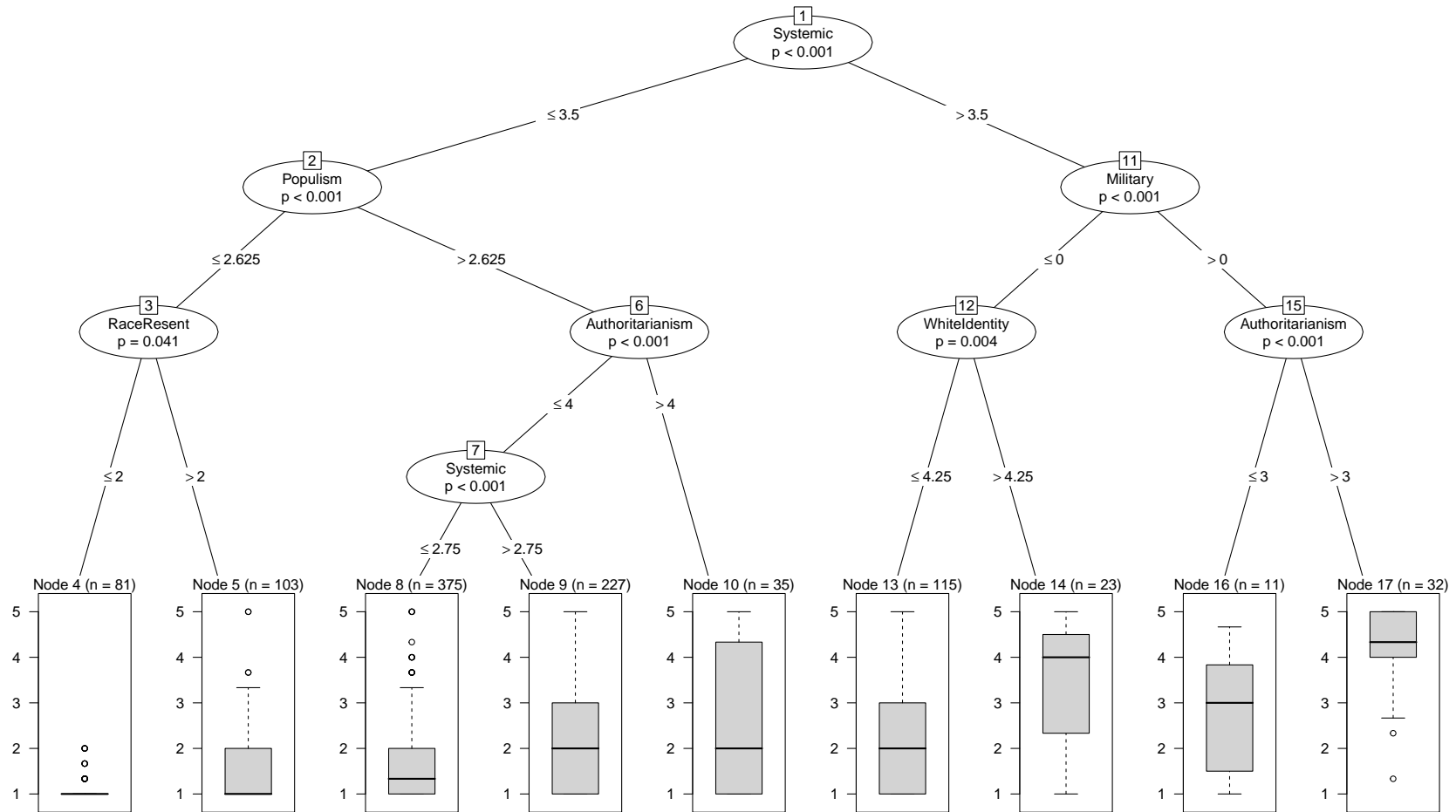
Six variables—systemic victimhood, populism, racial resentment, authoritarianism, current or former military enlistment, and white identity—prove useful in classifying levels of support for political violence. The interpretation proceeds just as above for all remaining nodes. As for the profiles that classify those high in support for political violence, we turn to the right side of the figure. Here, systemic victimhood, military service, white identity, and authoritarianism are the constellation of characteristics and orientations most predictive. Consider one high in systemic victimhood (>3.5), who is now or was previously in the military (>0), and is relatively low in authoritarianism (≤ 3). Such an individual would be classified into Node 16—about one standard deviation above the mean for support for violence. Should we consider a similar individual who is relatively high in authoritarianism (>3), they would be classified into Node 17—the 99th percentile of support for violence. Nodes 13 and 14 involve non-military personnel high in perceived systemic victimhood. Here, white identity proves to be additionally useful in classifying support for violence among such individuals. Those high in white identity are much more supportive of violence than their less white identified counterparts, despite having other similar characteristics. An ordinary linear model would hide this unique role of identity, which is couched within military status and perceptions of systemic victimhood.

Even though many other constructs and characteristics that we examined above correlate with support for violence, they simply do not exhibit as much unique predictive power as the 6 orientations pictured in Figure 6. On the one hand, we might expect as much. Indeed, many of the 24 correlates we examined are conceptually and empirically related to each other. On the other, there are several useful inferences to glean from the correlates that do prove uniquely predictive. First, they are all social-psychological orientations, with the exception of military service. This is in contrast to tangential literatures on political extremism and terrorism which sometimes argue that socioeconomic deprivation and social standing more generally—tangible social characteristics—promote (support for) political violence as a remedy for one’s troubles. Our results show that merely feeling like a victim

(e.g., systemic victimhood, white identity), blended with psychological postures that seek to resolve feelings of victimhood (e.g., support for authoritarian, populist, and racist ideas), can promote support for political violence.

Second, our results show that even though racial resentment and white identity, populism and authoritarianism overlap considerably, these constructs each play a distinct role in fostering support for violence. When it comes to racial attitudes, both in-group identities (i.e., white identity) and out-group orientations (i.e., racial resentment) matter—support for violence may be born of both in-group preservation and out-group denigration. Likewise, even though authoritarianism and populism are empirically related (Norris and Inglehart 2019), support for specific flavors of authority (i.e., authoritarianism) and a deep-seated mistrust of elites and the political establishment (populism) are also seemingly at logical odds in many ways. Our analysis shows that authoritarianism is mostly likely to lead to support for violence when one also subjectively feels like a victim—authoritarianism is, perhaps, used as a strong arm remedy for victimhood, rather than blind support for any authority figure.

Figure 6: Results of CART model. Distribution of support for political violence, for various combinations of independent variables, in the final nodes along horizontal axis.



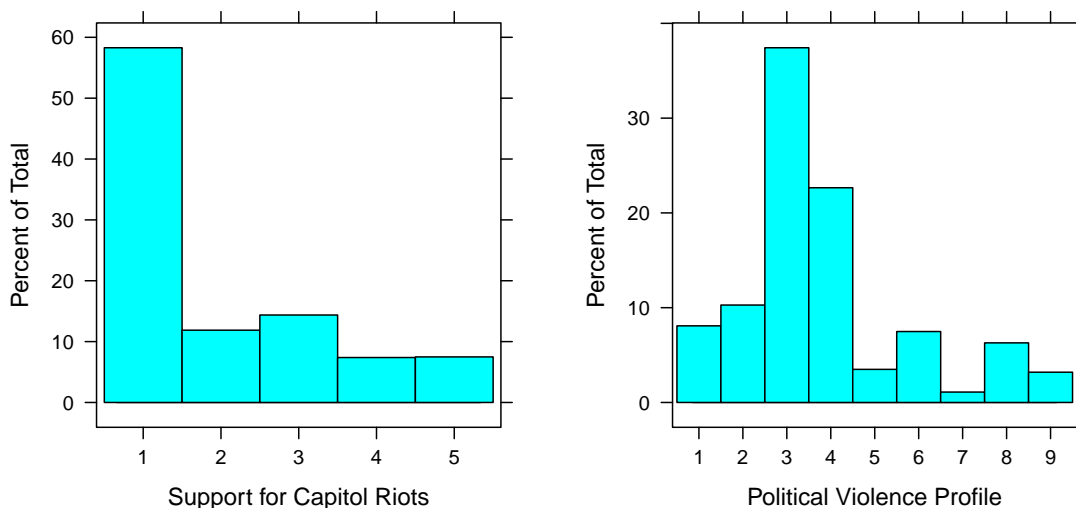
Explaining Attitudes about the Capitol Riots

Next, we examine how predictive the profile of characteristics identified by the CART model is when it comes to more specific non-normative attitudes. In particular, we seek to explain the extent to which Americans (dis)agree with the statement, “The riots at the U.S. Capitol building on January 6th were justified.” The distribution of responses appears in the lefthand panel of Figure 7 (where 5 corresponds to “strongly agree”). As with the more abstract sentiment about support for political violence explored above, fairly few individuals are supportive of the riot. Approximately 15% of Americans agree with the statement, 14% are neutral, and 71% disagree. These values accord with literature noting that support for violence is not particularly high (Westwood et al. 2021). In this instance, however, we know that it only takes a few individuals supportive of political violence to cause injury, death, and destruction, and to disrupt a critical democratic process.

To construct the political violence profile (PVP), we simply followed the nodes in Figure 6. There are 9 terminal nodes along the bottom of the figure, so our profile variable will have 9 categories. Those categorized at 9, the highest score on the PVP, scored greater than 3.5 on perceived systemic victimhood, 1 on military service, and greater than 3 on authoritarianism. Likewise, those categorized at 1, the lowest score on the PVP, scored less than or equal to 3.5 on perceived systemic victimhood, less than or equal to 2.625 on populism, and less than or equal to 2 on racial resentment.¹⁶ The distribution of the PVP appears in the righthand panel of Figure 7. Most people are classified into the 3 or 4 categories, reflecting low-neutral levels of perceived systemic victimhood, high-neutral levels of populism, and low-neutral levels of authoritarianism. This profile is uncorrelated with either partisan ($r = 0.001$, $p = 0.971$) or ideological ($r = -0.045$, $p = 0.192$) identities, though it is weakly correlated with the strength of ideological identities ($r = 0.082$, $p =$

¹⁶The white identity questions were only asked of white respondents. However, the conditional inference tree algorithm that we used is capable of handling missing data. Thus, all respondents were included in the CART. A very small subset of respondents (28 of 1,002) could be classified into either the 6 or 8 category in the PVP due missing values on white identity. Empirically, their classification into either category is immaterial—results are substantively and statistically identical.

Figure 7: Distribution of support for U.S. Capitol riots (left) and the violence profile (right).



0.017).

We take an iterative approach to explaining support for the Capitol riot. First, we estimate a baseline model, which includes partisan and ideological identities, as well as standard socio-demographic controls. Since rioters were present at the Capitol in order to attend Donald Trump’s “Save America” rally, we might expect that Republicans and conservatives are more likely to believe the riots were justified than Democrats and liberals. Next, we add to this model a measure of Trump support, operationalized using a 101-point thermometer of feelings about Trump. We expect this to be more strongly related to attitudes about the riot than either partisanship or ideology. To build off this, we next introduce the political violence profile to the model. Support for the Capitol riots is not merely an expression of intense partisanship or attachment to a political candidate, but an endorsement of the use of violent tactics in adjudicating political disagreements (to put it charitably). Thus, we expect the PVP to be strongly related to attitudes about the riot. We still expect support for Trump to relate to support for the Capitol riots, but that the coefficient magnitude will be attenuated when including the PVP. Finally, we suspect that there is an interactive relationship between Trump support and the PVP—the connection

between the PVP and attitudes about the riot should be strongest among people who exhibit the most positive feelings about Donald Trump.¹⁷

Each of these four models appear in Table 2.¹⁸ All independent variables have been rescaled to range from 0 to 1 in order to facilitate comparison of coefficients. In the first model, we find that, as expected, conservatives are more likely to believe the riots were justified, though we do not observe a relationship with partisanship. We also find that women and older individuals are less likely than men and younger people to believe the riots were justified. Moving to the second model, we see that Trump support is strongly related to attitudes about the riot. In the third model, we observe a coefficient for the PVP that is on par with (indeed, slightly greater than) that of Trump support. Moreover, the final model reveals a positive interactive relationship between Trump support and the PVP, precisely as we hypothesized.

To better understand the additive effects of the PVP and Trump support, we present model-predicted attitudes about the Capitol riot over the range of both variables in Figure 8.¹⁹ These two variables exhibit remarkably similar relationships. Among those lowest on the PVP and most negative about Donald Trump, the predicted level of support for riot is about a 1.5—somewhere between “strongly disagree” and “disagree.” Among those highest on the PVP and most positive about Donald Trump, the predicted level of support for the riot is about a 3, or “neutral.” While this may not seem particularly impressive, there are some caveats to keep in mind. First, even neutrality about political violence being justified in

¹⁷One may wonder why we do not simply include all of the potential correlates of political violence identified above in a model of support for the riot. The primary reason is that we know from the CART model that not all of the attitudes and orientations explored above are uniquely predictive of support for violence. As such, it would be difficult to derive expectations about which factors should be (most) influential in explaining support for the riot. That said, we estimated one large model analogous to the third model from Table 2, but also including every remaining correlate of political violence identified in Table 1. Even in this model, the PVP and Trump support exhibit the strongest relationship with support for the riot. Moreover, the interaction between Trump support and the PVP remains statistically significant ($p < 0.05$) in the interactive version of this model. This constitutes strong evidence for the predictive validity of the PVP. These model results appear in the appendix.

¹⁸Because of the skewed nature of the dependent variable, we also estimated these models using negative binomial regression. Results appear in the appendix.

¹⁹All other model variables are held at their mean in generating these predictions.

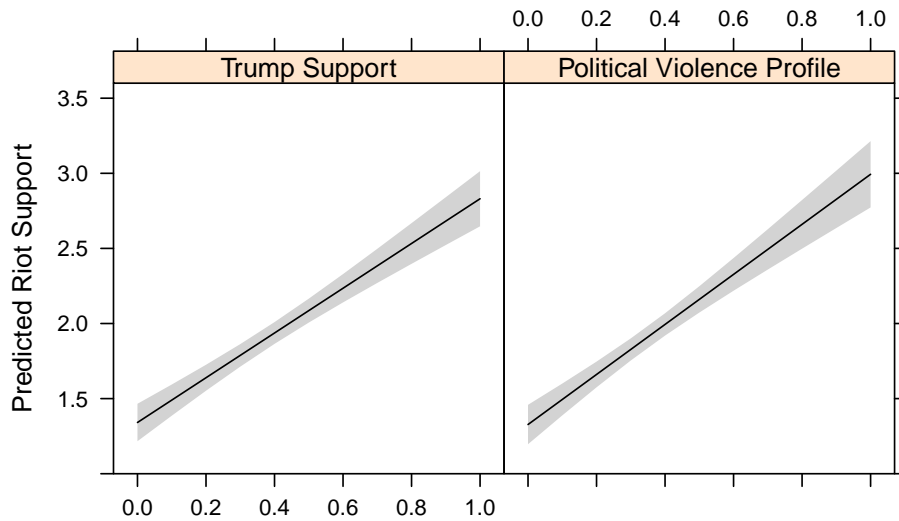
Table 2: OLS regressions of support for U.S. Capitol riots.

	(1)	(2)	(3)	(4)
Republican (vs. Democrat)	0.275*	-0.461***	-0.432***	-0.404**
	(0.124)	(0.135)	(0.127)	(0.126)
Independent (vs. Democrat)	0.667**	0.211	0.163	0.204
	(0.205)	(0.197)	(0.185)	(0.185)
Conservative (vs. Liberal)	0.231	-0.103	-0.051	0.005
	(0.133)	(0.128)	(0.121)	(0.122)
Moderate (vs. Liberal)	0.481**	0.232	0.216	0.234
	(0.178)	(0.169)	(0.159)	(0.158)
Partisan Strength	0.776***	0.518**	0.502**	0.522**
	(0.193)	(0.183)	(0.172)	(0.171)
Ideological Strength	0.672**	0.399*	0.289	0.292
	(0.215)	(0.202)	(0.191)	(0.190)
Interest in Politics	-0.147	-0.190	-0.220	-0.248
	(0.159)	(0.149)	(0.140)	(0.140)
Education	0.616***	0.601***	0.512**	0.476**
	(0.183)	(0.171)	(0.161)	(0.161)
Age	-2.062***	-1.863***	-1.359***	-1.416***
	(0.201)	(0.189)	(0.185)	(0.185)
Income	-0.150	-0.184	-0.053	-0.070
	(0.157)	(0.147)	(0.139)	(0.138)
Female	-0.340***	-0.311***	-0.220**	-0.231**
	(0.086)	(0.081)	(0.076)	(0.076)
White	0.085	0.040	0.152	0.136
	(0.123)	(0.115)	(0.109)	(0.109)
South	0.001	-0.047	-0.057	-0.050
	(0.088)	(0.083)	(0.078)	(0.077)
Trump Support		1.614***	1.427***	0.948***
		(0.151)	(0.143)	(0.220)
Political Violence Profile			1.659***	1.237***
			(0.162)	(0.219)
Trump Support × PVP				1.215**
				(0.426)
Constant	1.367***	1.626***	0.823***	1.014***
	(0.260)	(0.245)	(0.243)	(0.251)
R^2	0.189	0.290	0.373	0.379
n	815	815	815	815

Note: OLS coefficients with standard errors in parentheses

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

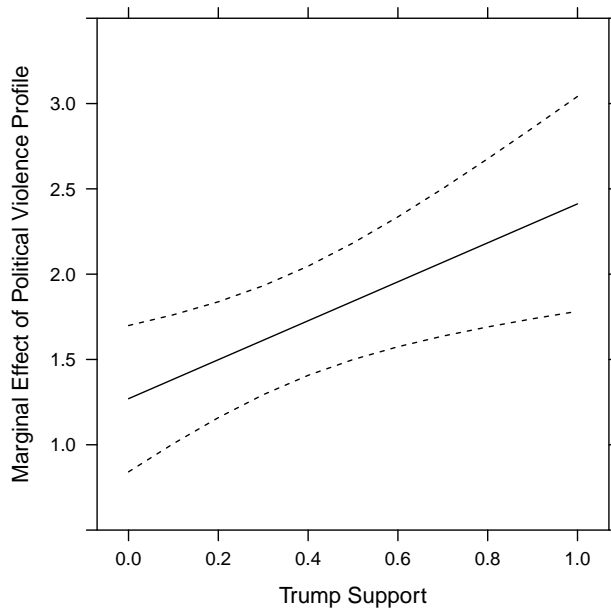
Figure 8: Predicted support for the U.S. Capitol riots by level of Trump support and violence profile, holding other variables at their mean. Bands represent 95% confidence intervals.



the context of lies about the integrity of an election is disconcerting. Second, these variables exhibit quite strong relationships, regardless of predicted levels of support—simply put, not many Americans admit to believing the riots were justified. Finally, the combination of Trump support and the PVP is likely to explain attitudes about the riot even better than either factor on its own.

Indeed, we do observe a significant interaction between Trump support and the PVP. The marginal effect of the PVP over the range of possible feelings about Trump appears in Figure 9. Even among those with the most negative feelings about Trump we observe a significant relationship between the PVP and support for the riot, as we might expect given the results of the additive model. This relationship significantly increases in strength as affect toward Trump becomes more positive. For those exhibiting the most positive feelings toward Trump (a 100 on the thermometer, which is 115 individuals in our sample, 29% of Republicans) and the highest score on the PVP, predicted support for the riots is 4.47 (95% CI: 4.06, 4.89), which is between “agree” and “strongly agree.”

Figure 9: Marginal effect of violence profile on support for the U.S. Capitol riots conditional on Trump support. Bands represent 95% confidence intervals.



This pattern is congruent with what others observed over the course of the Trump presidency. Donald Trump is particularly adept at recognizing, activating, and directing various social-psychological orientations and specific attitudes that other politicians have either ignored or only unsuccessfully manipulated, whether it be feelings of victimhood (Armaly and Enders 2021), authoritarianism (Knuckey and Hassan 2020), status threat (Mutz 2018), or tendencies toward violence. It is the combination of deep-seated commitment to Donald Trump—not merely Republican Party support or conservatism—and the willingness to engage in, or at least support the use of, political violence that results in a toxic blend of orientations strong enough to support hostility, ignore the disruption of a critical democratic process, and even spur action in some cases.

Conclusion

Our results indicate that support for political violence is strongly associated with a wide variety of psychological, political, and social factors, well beyond political identities, religion, and socioeconomic status, as identified by previous research. Moreover, with the exception of military service, it is the psychological orientations toward power, authority, and racial groups that appear to be the most predictive—in interaction with each other—of support for political violence. The individuals most supportive of political violence in our study exhibited high levels of perceived victimhood, authoritarianism, and white identity, as well as past or present military service. Importantly, this profile of characteristics is capable of explaining attitudes about real world events: especially when combined with strong levels of Trump support, the political violence profile provides the best explanation of who is most likely to believe the riots were justified. Beyond just the Capitol riot, identifying psychological and social factors that relate to support for violence can inform the correlates that may underlie future violence.

These patterns constitute an important link between individual political psychology and actual, behavioral engagement in destructive political acts, such as violence, by demonstrating which individual-level factors translate into support for violence. Reassuringly, neither Democrats nor Republicans, liberals nor conservatives appear to be asymmetrically supportive of political violence in the abstract; this is not *inherently* a partisan or ideological problem. However, feelings of victimhood and authoritarian personalities are hardly marginal orientations among the American mass public, and politicians—like Donald Trump—have seemingly recognized some utility in activating these orientations, presumably with a goal of expanding and mobilizing their base. Thus, better understanding which of the orientations underwriting support for political violence can be inflamed by elite cues, and how, may help in preventing physical outbreaks of political violence. While we do not ascribe causality in this paper, we do note that some of the factors we consider are “closer” in the likely causal

chain than others (e.g., some psychological elements may be a function of identity or sociodemographic characteristics). We encourage future research to disentangle the antecedents of the individual factors we identify.

This point is especially critical given that support for political violence appears to be fostered more by subjective social-psychological orientations (e.g., feelings of victimhood, racial resentment) than objective social conditions, such as socioeconomic deprivation or social standing, more generally. In other words, even though some individuals may become radicalized by poor economic conditions or an oppressive political or economic system, we find that more subjective orientations—such as merely *feeling* like a victim—are more predictive. This makes the etiology of constructs such as authoritarianism, perceived victimhood, and in- and out-group racial orientations all the more important to understand. There are many paths that may lead one to loathe the political establishment and other political enemies—many of them seem capable of fostering extremist attitudes regarding acceptable methods of interacting with that establishment and those enemies.

Our study is not without limitations. While we have good reason to examine the support for political violence in the U.S. after the riot of January 6, 2021, some of our results may be context-dependent—especially those involving uniquely American political identities, or those which may be impacted by “top down” elite communications and behavior. As such, we encourage others to replicate and extend this analysis in other political and cultural contexts. We also acknowledge that the link between attitudinal support for political violence and behavioral engagement in political violence is difficult to empirically decipher, let alone to infer from attitudinal measures alone. Indeed, some support for violence may even amount to nothing more than expressive responding—the intentional misreporting of true beliefs in an effort to express displeasure with the object of the question. That said, we did not include partisan figures/groups in any of the violence questions we employed, which should reduce any expressive responding that may occur. Moreover, even if support for violence does not translate into physical violence, support for breaking such a critical norm may still have

grave consequences for the health of democracy—surely attitudinal expressions of extremist sentiments such as this are worthy of systematic attention.

One additional limitation may be the use of an opt-in, non-probability sample, which have the potential to bias results (see Peyton, Huber and Coppock 2020; Westwood et al. 2021). As Coppock and McClellan (2019) and Strange et al. (2019) indicate, the use of such samples, while carrying some risk, is appropriate in many contexts and we have taken several steps to ensure the quality of our data, as described above. Nevertheless, we encourage increased scholarly focus on support for political violence using different measurements, research designs, data collection procedures, and platforms. Relatedly, we note that concerns common to survey research take on increased importance when attempting to measure something as substantively and normatively important as support for violence. Two concerns in particular—acquiescence bias/satisficing and measurement error—may impact inferences. We direct interested readers to the appendix, where we consider acquiescence bias and find little cause for concern. As for measurement error, we note (in Table 1) that all of our multi-item scales are statistically reliable. Thus, these common concerns do not appear to impact our analysis beyond what is typical. Still, we encourage future work to replicate and extend our analyses using different measures to ensure the robustness of findings.

Finally, we offer some practical advice for researchers studying support for political violence. As we demonstrate, and as others have noted (see Westwood et al. 2021), measuring support for political violence is inherently complicated. Yet, we are able to identify a small collection of psychological and identity factors that play into this support. We do not believe that researchers need to measure the same 24 potential correlates that we identify and conduct their own classification analysis, as the specific construction of the PVP will, to some degree, be sample-dependent. Rather, we think it is important for researchers to consider the correlates we have identified and use them in their analysis, either by combining them into a political violence profile as we have (i.e., using the classification cutoffs produced by the CART) or with other dimensional reduction techniques (such as principal component

analysis or factor analysis).²⁰ It is the constellation of factors—rather than each factor, individually—that is critical when making inferences about support for political violence; treating each variable individually may mask important nuance that is revealed when treating them as constitutive components of a profile of political violence.

The January 6 Capitol riots likely made political violence a less abstract, more tangible concept for many Americans who have never been threatened or seen someone else threatened with violence for political reasons. Even though forecasting individual acts of violence is an extremely difficult and largely empirically fruitless endeavor, understanding who tends to support the use of political violence—both generally and in relation to specific past events or future scenarios—can aid scholars and even law enforcement in preventing acts of violence, on the one hand, and fostering support for civil political interactions and institutions, on the other. As our results suggest, recognizing that groups of primarily psychological factors—rather than any single characteristic—likely underscore support for violence may be an important first step.

²⁰Analyses presented in the appendix indicate that measuring the political violence profile using these techniques produce results that are statistically similar, albeit substantively weaker, compared to those in Table 2.

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Supplemental Material for

“Who Supports Political Violence?”

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1 Variable question wording

Violence, developed by Uscinski and Parent 2014 and Kalmoe and Mason 2018, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- It is acceptable to use violence in advancing political goals these days.
- Violence is sometimes an acceptable way for Americans to express their disagreement with the government.
- Violence is justified if the members of the other side act violently first.

Systemic victimhood, developed by Armaly and Enders 2021, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- The system works against people like me.
- I feel that the world is “doing it” to me and that there is nothing I can do about it.
- The world is out to get me.
- The system is rigged to benefit a select few.

Egocentric victimhood, developed by Armaly and Enders 2021, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- I rarely get what I deserve in life.
- Great things never come to me.
- I usually have to settle for less.
- I never seem to get an extra break.

Conspiratorial thinking, developed by Uscinski and Parent 2014, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- Unseen patterns and secret activities can be found everywhere in politics.
- Much of our lives are being controlled by plots hatched in secret places.
- Even though we live in a democracy, a few people will always run things anyway.
- The people who really “run” the country, are not known to the voters.

Racial resentment, developed by Kinder and Sanders 1996, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- Irish, Italians, Jewish and many other minorities overcame prejudice and worked their way up. Blacks should do the same without any special favors.

- Generations of slavery and discrimination have created conditions that make it difficult for blacks to work their way out of the lower class.
- Over the past few years, blacks have gotten less than they deserve.
- It's really a matter of some people not trying hard enough; if blacks would only try harder they could be just as well off as whites.

Anxiety, developed by Spitzer et al. 2006, “Over the last 2 weeks, how often have you been bothered by the following problems?” all on four-point “not at all” to “nearly every day” scale (responses averaged across items):

- Feeling nervous, anxious or on edge
- Not being able to stop or control worrying
- Worrying too much about different things
- Trouble relaxing
- Being so restless that it's hard to sit still
- Becoming easily annoyed or irritable
- Feeling afraid, as if something awful might happen

Powerlessness, developed by Jolley and Douglas 2014, “How often are the following true?” all on five-point “never” to “always” scale (responses averaged across items):

- If you try hard enough, how often do you know what to do to overcome a problem?
- How often can you manage things you set out to do?
- How often do you feel helpless?

Populism, developed by Oliver and Rahn 2016, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- People like me don't have much say in what government does.
- Politics usually boils down to a struggle between the people and the powerful.
- The system is stacked against people like me.
- It doesn't really matter who you vote for because the rich control both political parties.
- People at the top usually get there from some unfair advantage.
- I'd rather put my trust in the wisdom of ordinary people than the opinions of experts and intellectuals.

- When it comes to really important questions, scientific facts don't help very much.
- Politics is ultimately a struggle between good and evil.

White Identity, developed by Jardina 2019, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- Being white is important to my identity.
- White people in this country have a lot to be proud of.
- Whites in this country have a lot in common with one another.
- It is important that whites work together to change laws that are unfair to whites.

Authoritarianism, developed by Heller et al. 2020, all on five-point “strongly disagree” to “strongly agree” scale (responses averaged across items):

- Established conducts should not be questioned.
- People should leave important decisions to those in charge/the leaders.
- Troublemakers should clearly feel the effects of the fact that they are unwanted in society.

Republican: (derived from standard two-part branching question measure)

0. Democrat
1. Republican

Partisan strength: (derived from standard two-part branching questions)

1. Independent
2. Leaning
3. Weak identification
4. Strong identification

Conservative: (derived from “We hear a lot of talk these days about liberals and conservatives. When it comes to politics, how do you think about yourself?”)

0. Liberal
1. Conservative

Ideological strength: (derived from, “We hear a lot of talk these days about liberals and conservatives. When it comes to politics, how do you think about yourself?”)

1. Moderate
2. Leaning
3. Weak identification
4. Strong identification

Perceived corruption: “How many of the people running the government would you say are corrupt?”

1. None
2. A few
3. About half
4. Most
5. All

Government trust: “How often can you trust the federal government in Washington to do what is right?”

1. Never
2. Sometimes
3. About half the time
4. Most of the time
5. Always

Interest in politics: “How interested would you say you are in politics?”

1. Not at all interested
2. Not very interested
3. Somewhat interested
4. Very interested

Religiosity: “Do you go to religious services every week, almost every week, once or twice a month, a few times a year, or never?”

1. Never
2. A few times a year

3. Once or twice a month
4. Almost every week
5. Every week

Evangelical: “Would you describe yourself as a born-again or evangelical Christian, or not?”

0. No, not born-again or evangelical Christian
1. Yes, born-again or evangelical Christian

Health Insurance: “Do you currently have health insurance?”

0. No
1. Yes

Military personnel: “Do you currently, or have you ever, served in the U.S. military?”

0. No, I have never served
1. Yes, I currently serve OR Yes, I served in the past

Income: “What was your total household income in the past 12 months?”

1. Under \$15,000
2. \$15,000 to \$24,999
3. \$25,000 to \$49,999
4. \$50,000 to \$99,999
5. \$100,000 and above

Education: “What is the highest level of school you have completed or the highest degree you have received?”

1. Did not graduate high school
2. High school diploma or equivalent (GED)
3. Some college but no degree
4. Bachelor’s degree (BA, BS)
5. Graduate degree (MA, MS, MD, PhD)

2 Sample characteristics

Our sample was fielded by Lucid, which is a survey marketplace that recruits respondents from dozens of sample suppliers. While the sample is not a probability sample, the demographic characteristics of respondents match those of the broader population according to U.S. Census data and the platform has been found to perform very well in generating accurate estimates of various political attitudes. Coppock and McClellan [?], for example, highlight the utility of opt-in, online platforms, writing “...it is the rare theory whose scope conditions specifically exclude the sort of people who take online surveys...”. Strange et al. [?] call opt-in, online subject pools “quality frontiers of human subjects research” and note that extensive study has “demonstrated the viability of paid online subjects for social and behavioral science research.” This paper also highlights differences between paid subjects and volunteers, and finds that paid subjects pay more attention to many aspects of the survey.

We also highlight that Lucid conducts several quality checks on respondents before they make it into the pool and are allowed to complete new surveys. From the Lucid website:

- We block bots using multiple third-party services, such as Google reCAPTCHA.
- We screen every Lucid Theorem participant with attention-checker and open-ended questions, using machine learning to screen out bad participants.
- As a marketplace, Lucid partners with its sample suppliers to have partners implement state-of-the-art fraud prevention and quality detection services.

For additional examples of published academic work that uses Lucid, see the following: <https://luc.id/citations/>.

Table 1: Demographic characteristics from 2021 sample compared to 2019 American Community Survey (i.e., most recently available census data).

Characteristic (range)	2021	Census Estimate
Age (18-86)	46*	38.1
Age 18-24 (%)	12.48	13
Age 25-44 (%)	34.23	35
Age 45-64 (%)	33.73	34.8
Age >65 (%)	19.56	17.1
<\$14,999 (%)	15.17	9.8
\$15,000-24,999 (%)	9.88	8.3
\$25,000-49,999 (%)	24.65	20.3
\$50,000-99,999 (%)	26.15	30.2
\$100,000-199,999 (%)	15.07	22.9
>\$200,000 (%)	3.40	8.5
Female (%)	52.79	50.8
Party ID (1-7)	3.65	3.89**
Democrat (%)	49.27	46.53**
Independent (%)	15.52	11.74**
Republican (%)	35.21	41.73**
Ideology (1-7)	4.10	4.09**
Liberal (%)	32.87	35.39**
Moderate (%)	30.56	25.77**
Conservative (%)	36.57	38.86**
Race:		
White (%)	78.24	72.5
Black (%)	11.58	12.7
Hispanic (%)	10.87	18
Other (%)	3.89	5.9
Region:		
Northeast (%)	19.96	17.2
Midwest (%)	19.06	20.9
South (%)	39.12	38.1
West (%)	21.85	23.8

* Median used.

** Value from 2020 ANES.

3 Investigation of potential satisficing

In this section, we undertake an investigation of potential satisficing among respondents in order to ensure that 1) our estimates of support for political violence and other constructs are not inflated and that 2) any potential satisficing is not occurring in the same “direction” across independent and dependent variables, thereby inflating correlations.

First, we had to create a measure of (potential) satisficing. For each variable that is operationalized vis-à-vis a multiple-item scale, we computed the standard deviation, for each respondent, across the items belonging to the scale. A standard deviation of 0 signifies that the respondent provided the same answer to all questions. We note that this is only a very coarse measure of *potential* satisficing—respondents are obviously “allowed” to possess similar attitudes across the various items. Indeed, we would expect this to be true if the indicators were “true” parallel measures, as the summated rating model assumes them to be when items are summed or averaged into a scale/index. In any case, we are certainly over-estimating satisficing with this operationalization.

To examine the potential impact of satisficing, we first present, in Table 2, the average of each scale among potential satisficers (those with a standard deviation of 0 across all scale items) and others (those with a standard deviation greater than 0, meaning that a different answer was provided for at least one question in the scale). There are two potential causes of worry: 1) inflating our estimates of some of these variables (especially support for political violence) and 2) satisficing happening in the same “direction” across items, thereby inflating correlations between constructs. We find support for neither notion in Table 2. First, potential satisficers register significantly greater levels of only 2 of the 10 scales (one of which was identified by the CART procedure, white identity). We find that for 5 of the 10 variables—including support for political violence—potential satisficers registered significantly lower levels. There’s no difference between potential satisficers and others with respect to the 3 remaining variables, all of which were identified by the CART model as important predictors of support for violence. Thus, we do not appear to be systematically over-estimating the levels of support for political violence and other constructs because of satisficing, nor does any potential satisficing appear to be exhibiting a consistent effect across scales.

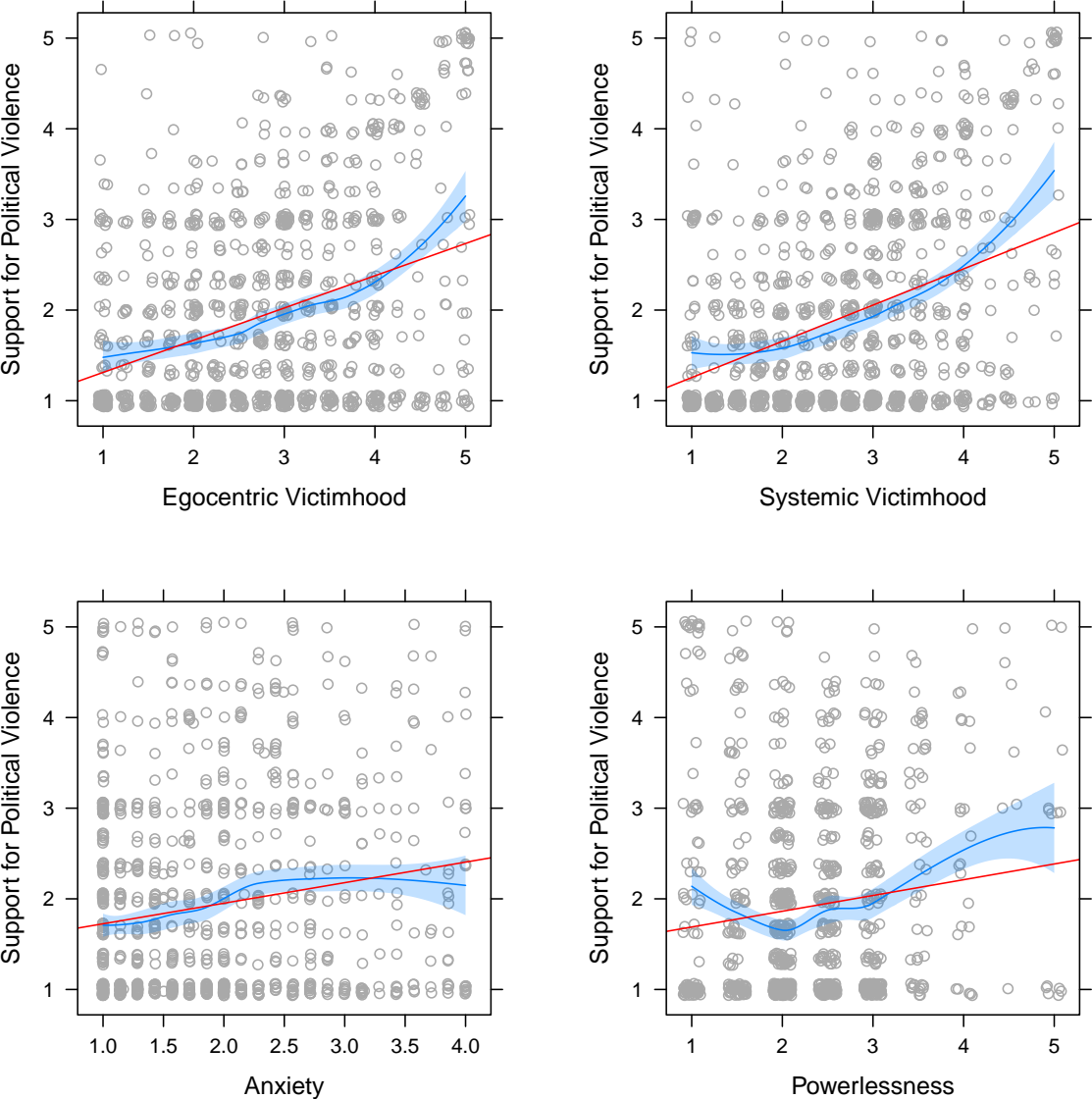
We also considered the potential correlation between satisficing and our dependent variables. To create a general measure of satisficing, we first classified respondents as potential satisficers (coded 1) or not (coded 0) for with respect to each scale in Table 2 except for support for political violence. Then we summed across these 9 variables to create a count that ranges from 0 (no satisficing for any scale; i.e., standard deviations greater than 0 across the board) to 9 (potential satisficing in each scale; i.e., standard deviations of 0 across the board). The correlation between this measure and support for political violence is 0.037 ($p=0.244$); the correlation with support for the Capitol riot is 0.015 ($p=0.628$). In other words, potential satisficing does not predict support for either dependent variable.

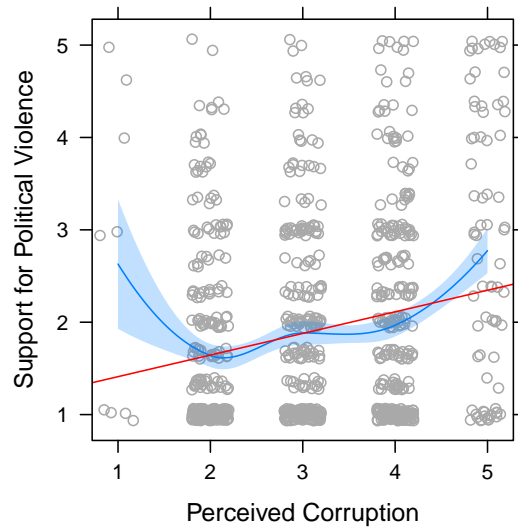
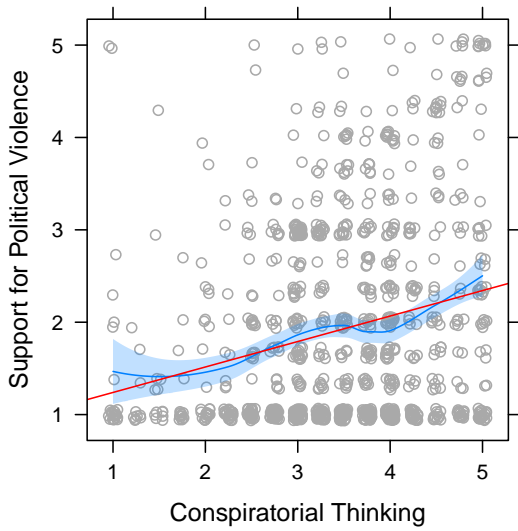
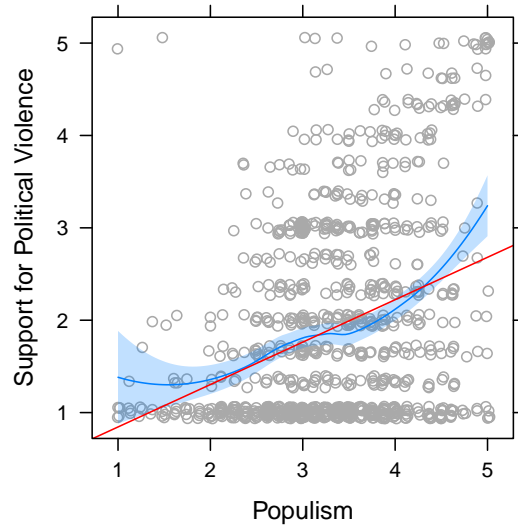
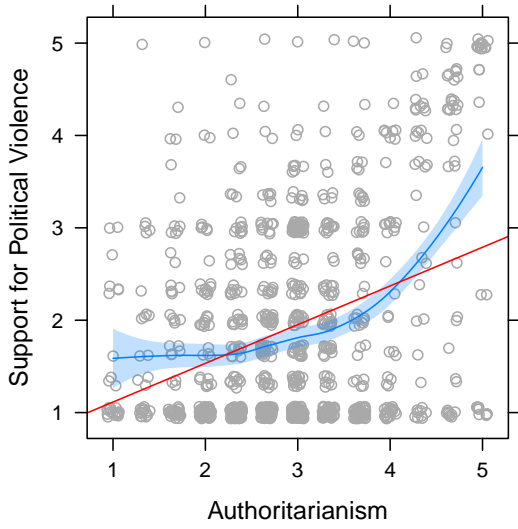
Table 2: Scale averages for potential satisficers and everyone else.

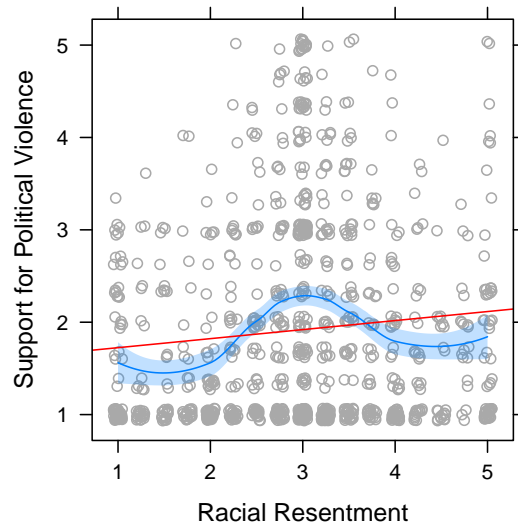
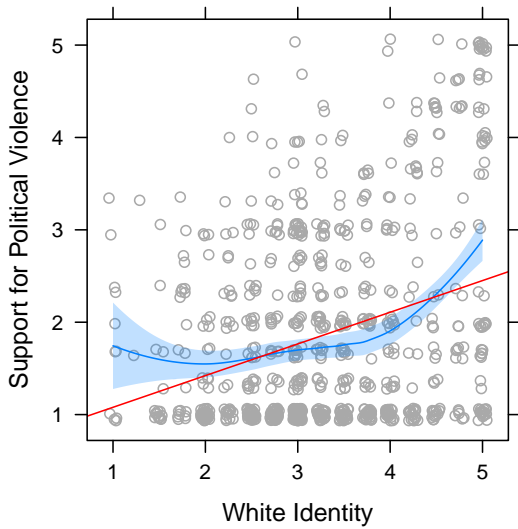
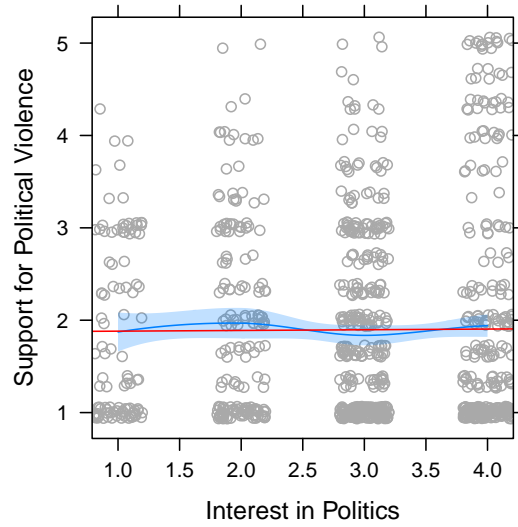
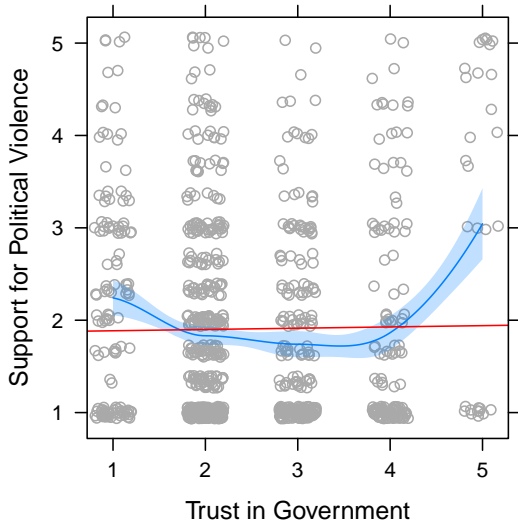
	Potential Satisficer	Unlikely Satisficer	<i>p</i> -value for difference
Support Violence	1.55	2.43	<0.001
Populism	3.44	3.32	0.21
Authoritarianism	2.93	2.90	0.65
White Identity	3.62	3.18	<0.001
Racial Resentment	2.91	2.96	0.51
Systemic Victimhood	2.29	2.74	<0.001
Egocentric Victimhood	2.47	2.80	<0.001
Anxiety	1.39	2.04	<0.001
Powerlessness	2.24	2.39	0.003
Conspiratorial Thinking	3.66	3.38	<0.001

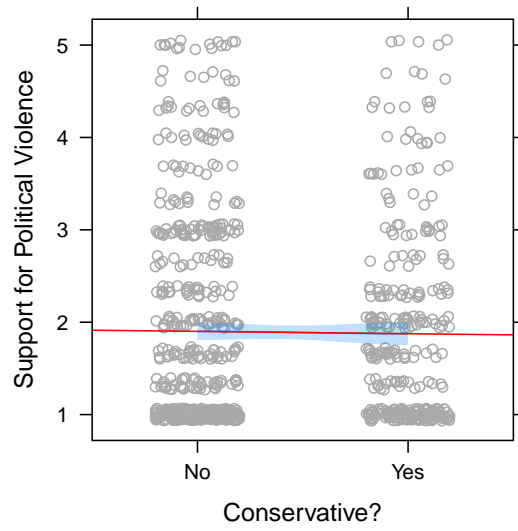
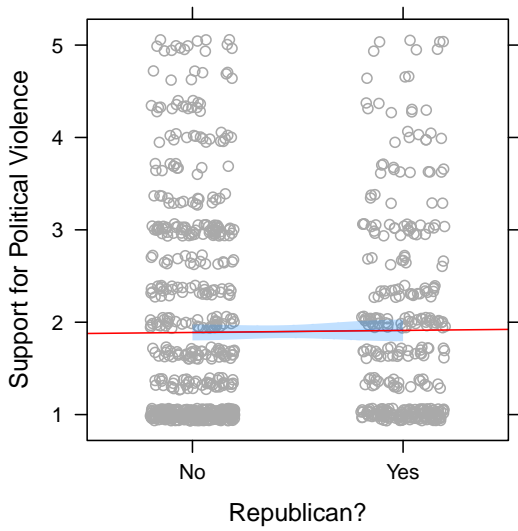
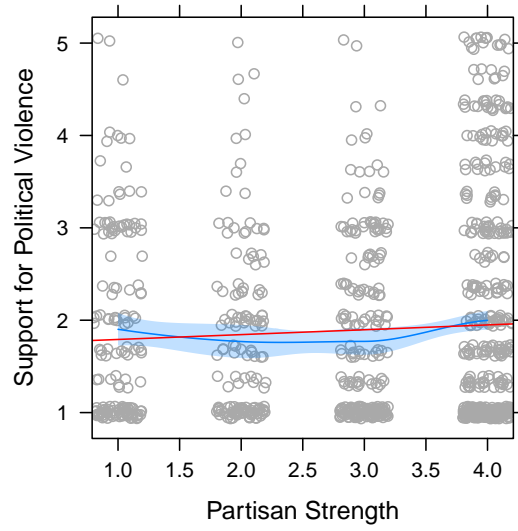
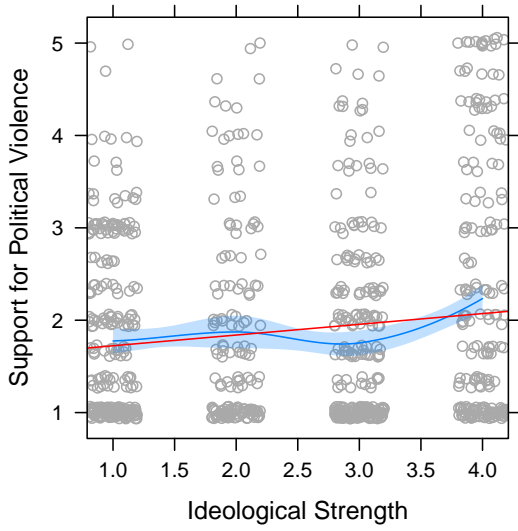
4 Relationship between support for violence and each independent variable

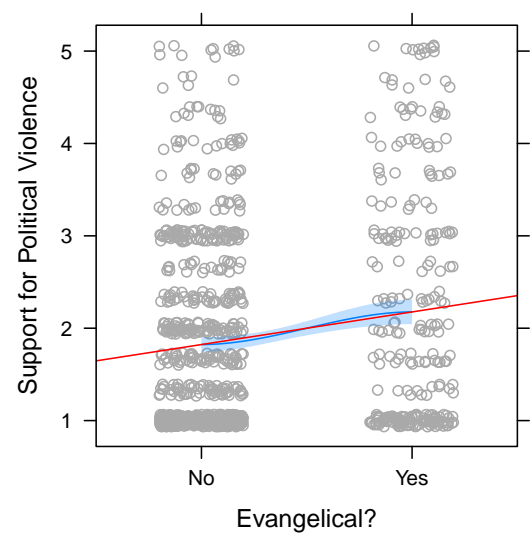
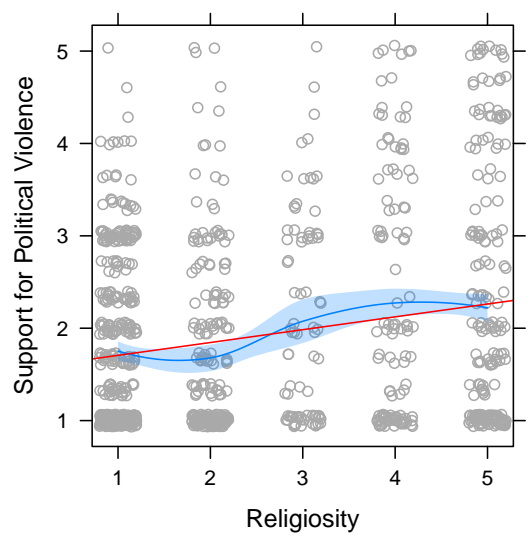
Figure 1: Scatterplots of relationships between support for violence and all substantive independent variables of interest. LOWESS curves with 95% confidence intervals in blue; OLS fit lines in red.











5 Negative binomial model estimates

Figure 2: Predicted support for the U.S. Capitol riots by level of Trump support and violence profile, holding other variables at their mean. Bands represent 95% confidence intervals. Negative binomial model.

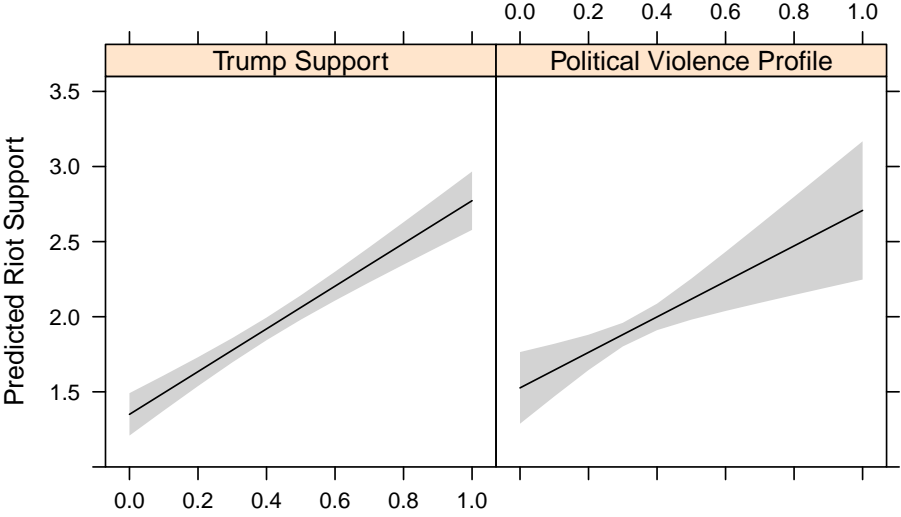


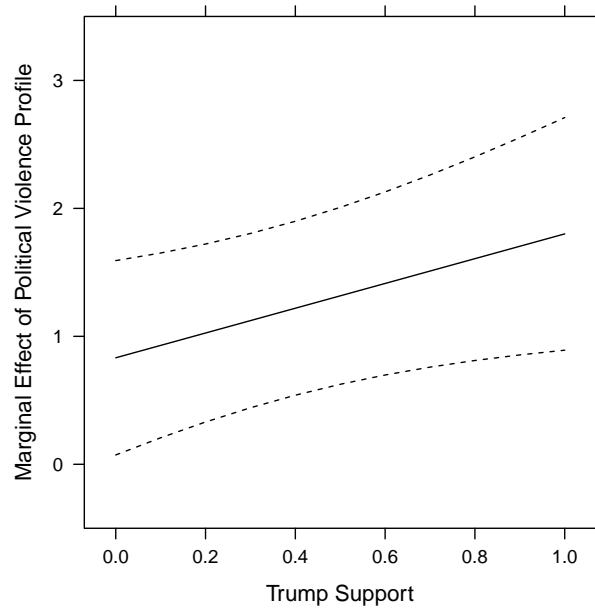
Table 3: Negative binomial regression models of support for the U.S. Capitol riot.

	(1)	(2)	(3)	(4)
Independent (vs. Democrat)	0.376** (0.132)	0.121 (0.135)	0.130 (0.135)	0.123 (0.135)
Republican (vs. Democrat)	0.155* (0.073)	-0.213* (0.086)	-0.176* (0.084)	-0.180* (0.085)
Moderate (vs. Liberal)	0.253* (0.112)	0.114 (0.112)	0.116 (0.112)	0.114 (0.112)
Conservative (vs. Liberal)	0.119 (0.078)	-0.043 (0.081)	0.021 (0.080)	0.015 (0.081)
Partisan Strength	0.413*** (0.123)	0.271* (0.123)	0.272* (0.124)	0.268* (0.124)
Ideological Strength	0.357** (0.129)	0.206 (0.131)	0.128 (0.131)	0.128 (0.131)
Interest in Politics	-0.091 (0.096)	-0.115 (0.096)	-0.150 (0.096)	-0.148 (0.096)
Education	0.332** (0.110)	0.312** (0.110)	0.238* (0.111)	0.242* (0.111)
Age	-1.102*** (0.123)	-1.008*** (0.125)	-0.803*** (0.130)	-0.798*** (0.131)
Income	-0.080 (0.095)	-0.115 (0.095)	-0.069 (0.095)	-0.067 (0.096)
Female	-0.181*** (0.053)	-0.161** (0.053)	-0.118* (0.054)	-0.117* (0.054)
White	0.030 (0.074)	0.008 (0.074)	0.059 (0.075)	0.061 (0.075)
South	0.009 (0.053)	-0.014 (0.053)	-0.015 (0.053)	-0.015 (0.053)
Trump Support		0.808*** (0.095)	0.692*** (0.094)	0.745*** (0.151)
Political Violence Profile			0.742*** (0.102)	0.792*** (0.150)
Trump Support × PVP				-0.114 (0.254)
Constant	0.328* (0.161)	0.474** (0.161)	0.141 (0.167)	0.116 (0.175)
Pseudo- R^2	0.054	0.082	0.102	0.102
n	815	815	815	815

Note: MLE coefficients with standard errors in parentheses.

* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Figure 3: Marginal effect of violence profile on support for the U.S. Capitol riots conditional on Trump support. Bands represent 95% confidence intervals. Negative binomial model.



6 Garbage can model estimates

Table 4: OLS regression models of support for the U.S. Capitol riot including all correlates of support for violence.

	(1)	(2)	(3)	(4)
Partisanship	0.273 (0.153)	-0.396* (0.160)	-0.415** (0.159)	-0.396* (0.159)
Ideology	-0.104 (0.194)	-0.403* (0.185)	-0.363* (0.184)	-0.312 (0.185)
Interest in Politics	0.087 (0.155)	-0.019 (0.146)	-0.032 (0.145)	-0.033 (0.144)
College	0.100 (0.092)	0.112 (0.086)	0.093 (0.086)	0.084 (0.086)
Age	-1.569*** (0.223)	-1.372*** (0.210)	-1.305*** (0.210)	-1.379*** (0.212)
Income	0.222 (0.165)	0.169 (0.155)	0.147 (0.154)	0.137 (0.154)
Female	-0.056 (0.092)	-0.068 (0.087)	-0.082 (0.086)	-0.096 (0.086)
Residence in the South	-0.016 (0.089)	-0.043 (0.084)	-0.046 (0.083)	-0.043 (0.083)
Egocentric Victimhood	0.168** (0.058)	0.170** (0.055)	0.158** (0.054)	0.154** (0.054)
Systemic Victimhood	0.057 (0.063)	0.048 (0.059)	-0.076 (0.069)	-0.087 (0.069)
Anxiety	-0.061 (0.063)	-0.055 (0.059)	-0.066 (0.058)	-0.078 (0.059)
Powerlessness	-0.019 (0.063)	0.017 (0.059)	0.014 (0.059)	0.030 (0.059)
Conspiracy Thinking	0.088 (0.062)	0.053 (0.059)	0.052 (0.058)	0.058 (0.058)
Perceived Corruption	0.167** (0.055)	0.132* (0.052)	0.116* (0.052)	0.116* (0.052)
Trust in Government	0.130* (0.051)	0.116* (0.048)	0.104* (0.047)	0.099* (0.047)
Populism	0.038 (0.083)	0.022 (0.078)	-0.075 (0.083)	-0.071 (0.082)
Authoritarianism	0.243*** (0.056)	0.248*** (0.053)	0.199*** (0.054)	0.196*** (0.054)
White Identity	0.301*** (0.055)	0.243*** (0.052)	0.230*** (0.052)	0.231*** (0.052)
Racial Resentment	0.097 (0.052)	0.013 (0.050)	-0.006 (0.050)	0.010 (0.050)
Health Insurance	-0.083 (0.169)	-0.067 (0.158)	-0.064 (0.157)	-0.063 (0.157)
Military Service	0.210 (0.121)	0.111 (0.114)	0.044 (0.115)	0.028 (0.115)
Trump Support		1.420*** (0.153)	1.422*** (0.152)	1.033*** (0.246)
Political Violence Profile			1.181*** (0.346)	0.832* (0.387)
Trump Support × PVP				0.968* (0.481)
Constant	-1.378** (0.420)	-0.750 (0.400)	-0.112 (0.438)	-0.014 (0.440)
R^2	0.401	0.474	0.484	0.487
n	644	644	644	644

Note: OLS coefficients with standard errors in parentheses.

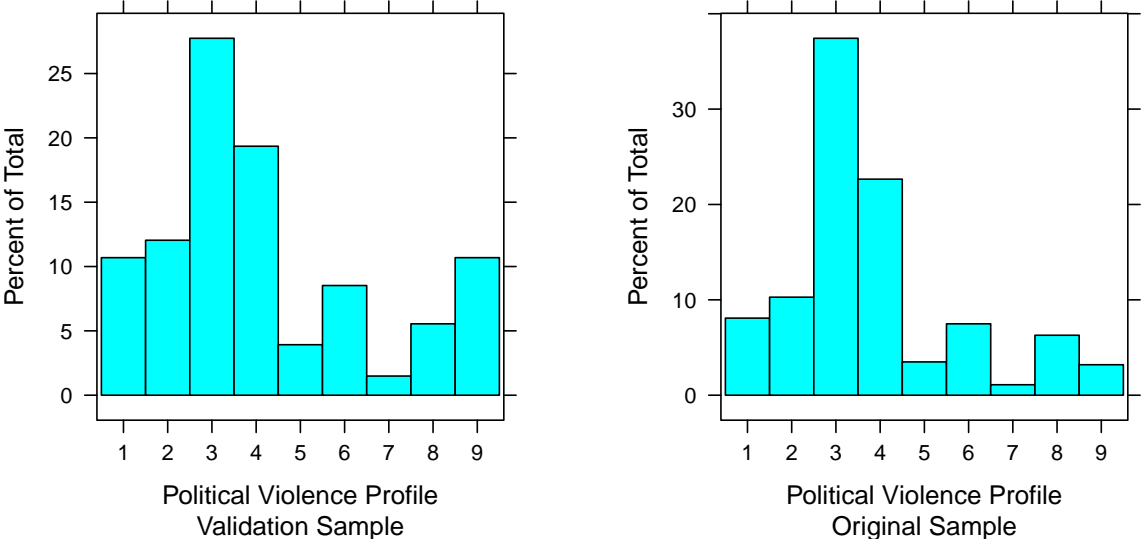
* $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

7 Validating the PVP with a new sample

One way to determine whether the model is identifying a general relationship is to use a separate dataset to generate the independent variable of interest without using the CART model and see if our conclusions are the same. We have done just that. Specifically, we asked 750 respondents all of the necessary survey items to reconstruct the regression models in Table 2 of the manuscript. We utilized MTurk for this task, and we acknowledge all of the issues that always arise with MTurk.¹ We are comfortable with MTurk for this task because we expect any bias induced by the platform will work against us.

Figure 4 displays the distribution of the PVP for both our validation sample and our original sample. They are remarkably similar. If our CART model was identifying a constellation of variables specific to our sample (i.e., peculiar and/or idiosyncratic correlations, non-linearities, etc.), it is unlikely that creating a profile from those variables would produce similar results across different samples. Thus, that we see a similar distribution when using the information from the CART model is reassuring.

Figure 4: Distribution of the violence profile for validation sample (left) and original sample (right).



Next, we used the PVP from the validation sample to replicate our regression models. In short, the estimates using our original sample look very similar to the estimates from the validation sample. We summarize the relevant coefficients in Table 5; note that these replication models take the exact form of those presented in Table 2 of the main text, even though we don't present the estimates associated with controls here.

This is all to state that the CART model produced a subset of variables and cut-off points that, when used in the construction of a variable, operates nearly identically

¹Just as in our Lucid survey, we removed respondents who failed one or both attention checks and those who rushed through the survey.

Table 5: Coefficients from OLS regressions of 1/6 justified onto Trump, PVP, and interaction. All coefficients are statistically significant at $p < 0.01$.

	Validation sample		Original sample	
Trump Support	1.46	0.99	1.43	0.95
PVP	1.97	1.41	1.66	1.24
Trump*PVP		1.21		1.22

across samples. While we are not willing to assert that we have identified the one-and-only, definitive general relationship, this serves as promising confirmatory evidence that we have uncovered a relationship that is sample independent.

8 Measuring the PVP with PCA and factor analysis

To demonstrate how other methods at constructing the PVP may be acceptable enough, we constructed the PVP using both PCA (retaining the first component scores) and factor analysis (predicting scores along the first factor; iterated principal axis factoring was the estimation method) and replicated the regressions from Table 2 in the manuscript. We do this using the MTurk sample described in the section above (again, to highlight that it is the variables selected by the CART, rather than the original Lucid sample, that is important). We present the pairwise correlations between each of the differently constructed PVP variables in Table 6, and show the relevant coefficients (including Trump support) in Table 7.

Each of these tables highlights that it does not matter a great deal how we construct our PVP variable—CART, PCA, and factor analysis all produce substantively similar results. However, the relationships are weaker when using the PCA or FA versions of the PVP. So, constructing our variable with the CART classification is not necessarily “correct,” but it performs better than the other versions, *even out of sample*. It seems to us that constructing the PVP using PCA or factor analysis is adequate, especially in instances where it is being used as a control. We have added guidance for researchers based on these analyses to the conclusion of the paper.

Table 6: Pairwise correlation between differently constructed PVP variables. All correlations significant at $p < 0.001$.

	PCA	FA
CART	0.775	0.785
FA	0.995	–

Table 7: Coefficients from OLS regressions of 1/6 justified onto Trump, PVP, and interaction using different constructions of PVP.

	CART		PCA		FA	
Trump Support	1.46*	0.99*	0.78*	0.79*	0.85*	0.85*
PVP	1.97*	1.41*	0.48*	0.51*	0.83*	0.85*
Trump Support \times PVP		1.21*		-0.06		-0.04

Note: * $p < 0.05$

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