

Let's Talk Politics

A Naive Approach for Measuring Political Sophistication*

Patrick W. Kraft[†]

Abstract

This paper proposes a simple but powerful framework to measure political sophistication based on open-ended survey responses. *Discursive sophistication* utilizes automated text analysis methods to capture the complexity of individual attitude expression. I validate the approach by comparing it to conventional political knowledge metrics in multiple studies using different batteries of open-ended items. The paper then illustrates how the measure can help refine previous insights from the literature such as the oft-cited gender gap in political knowledge. Women might know fewer facts about institutions and elites, but they do not differ substantively in the sophistication of their expressed political beliefs.

Keywords: political sophistication, open-ended responses, text analysis, gender gap

Word count: 5873 (via TeXcount)

*Previous versions of this manuscript have been presented at Polmeth 2016, MPSA 2017, EPSA 2017, ISPP 2017, and APSA 2017. I thank Talbot Andrews, Alexa Bankert, Jason Barabas, Scott Clifford, Andy Delton, Kathy Dolan, Peter DeScioli, Stanley Feldman, Fabrizio Gilardi, Bill Jacoby, Jennifer Jerit, Yanna Krupnikov, Jamie Monogan, Hannah Nam, the Nam Lab, Michael Peress, and Arthur Spirling for helpful comments on previous versions of this manuscript. Special thanks to Céline Colombo, Scott Clifford, and Jennifer Jerit for sharing their data. The manuscript and code are available on GitHub: <https://github.com/pwkraft/knowledge>.

[†]Assistant Professor, University of Wisconsin-Milwaukee, kraftp@uwm.edu.

One of the most important tasks for citizens in modern democracies is to vote for candidates who represent their interests and hold their elected officials accountable. While there have been longstanding debates about whether citizens are sufficiently informed to fulfill this task, fundamental issues regarding the measurement of knowledge continue to plague the discipline (Mondak, 2001; Sturgis, Allum, and Smith, 2008; Pietryka and MacIntosh, 2013). Most analyses rely on batteries that assess individuals' factual knowledge about political institutions and office-holders (e.g., Delli Carpini and Keeter, 1996). Theoretically, these survey questions should cover information that is necessary and/or sufficient for citizens to make competent decisions in a given context (c.f., Lupia, 2006, 2015). Yet, determining such a set of items proves to be extremely difficult, especially since there are systematic differences in types of knowledge (Barabas et al., 2014). Even within a given policy area, people may disagree about which facts are crucial for political competence due to inherent value differences (Lupia, 2015).

Despite these difficulties, most empirical studies rely on a set of off-the-shelf knowledge questions rather than justifying their choices theoretically. As Lupia (2006, 219) points out, “[m]ost political knowledge questions are not derived from a replicable or transparent logic about how their answers bear on a voter’s ability to make decisions of a particular quality.” It is therefore not surprising that conventional metrics do not properly capture policy-specific information (e.g., Gilens, 2001) or other knowledge relevant to citizens’ preferences and decision-making (c.f., Gruber, 2001, 43–68). In a recent review, Cramer and Toff (2017, 756) eloquently summarize: “All of this work suggests that we are missing a lot by equating information levels as measured in traditional knowledge batteries with civic competence. By focusing on what people do not know rather than what they do know and how they use that information, we are likely missing the empirical reality of citizens’ political knowledge.”

While some scholars contend that simple tests of factual information are nevertheless the best available proxy for political awareness (e.g., Zaller, 1990), others rely on broader conceptualizations of sophistication that incorporate additional dimensions such as education and income (e.g., Jacoby, 2006). Notwithstanding, the fundamental issue remains that knowledge quizzes rarely

cover information that is relevant for citizen competence (Lupia, 2006). In a similar vein, Druckman (2014) describes individual levels of political information as inadequate to measure “quality opinion” since there is no consensus about what information is necessary in the first place. Instead, Druckman advocates “*less focus on the content/substance* of opinions [...] and *more* on the *process* and specifically the *motivation* that underlies the formation of those opinions” (2014, 478, emphasis in the original).

The framework proposed herein follows this call by using a person’s open-ended responses to develop an indicator of *discursive sophistication*. The measure examines how respondents discuss their political beliefs in their own words and incorporates information about the number of considerations raised, the relative descriptiveness in word choice, as well as the level of opinionation. The approach is therefore *naive* in that it does not presuppose pieces of information as necessary for political competence but rather examines the respondents’ justification of their preferences at face value. Measuring sophistication based on people’s verbatim attitude expression provides two major advantages compared to off-the-shelf factual knowledge items: (1) it directly captures the extent to which a respondent’s political beliefs are based on elaborate reasoning, and (2) it can easily pinpoint competence in specific areas by incorporating targeted open-ended items.

I validate the measure across multiple data sets by comparing it to conventional factual knowledge scores as predictors of various indicators of competence. While the measures share a considerable amount of variance, they are far from equivalent. Indeed, discursive sophistication is a stronger predictor of turnout and other forms of political participation than traditional metrics. After validating the measurement approach, the paper illustrates how discursive sophistication can help refine previous insights in the literature by re-examining an oft-cited finding in empirical research—the gender gap in political knowledge. Contrary to previous research, I find no evidence for such a gap based on open-ended responses. While women might score lower than men on factual knowledge about political institutions and elites, there are no differences in the complexity of expressed political attitudes.

1 Opinion Formation and Attitude Expression

In modern democracies, citizens can engage in politics through various means such as voting in local, state, or federal elections. Depending on the institutional setup, they may also directly decide on specific policies through referenda. In these contexts, we are concerned with the ability of citizens to make high quality decisions in accordance with their underlying interests. Given that it is challenging to determine what information is indeed necessary and/or sufficient to make competent decisions in a given context, a useful alternative is to concentrate on whether people are motivated to engage in elaborate reasoning when forming their preferences (Druckman, 2014). In previous research, scholars have induced people to engage in in-depth processing by asking them to *justify* their opinions (e.g., by providing specific reasons; Kunda and Sinclair, 1999; Redlawsk, 2002; Bolsen, Druckman, and Cook, 2014a; Druckman, 2014). In an analogous way, we can examine *how* citizens justify their preferences in order to evaluate whether they engaged in elaborate and sophisticated reasoning (see also Rosenberg, 1988; Rosenberg, Ward, and Chilton, 1988). If respondents are motivated and able to engage in in-depth processing to form quality opinions, they should discuss multiple considerations related to a political issue and show awareness of arguments for and against certain positions (Cappella, Price, and Nir, 2002). Rather than trying to develop recall items that presupposes a set of facts as necessary for political competence, I therefore analyze *how* individuals discuss their preferences related to a given political task.¹

My approach is consistent with influential theoretical accounts of political sophistication which focus on the *structure* of belief systems. For example, Converse (1964) emphasizes the importance of the level of conceptualization as the main characteristic of sophistication rather than isolated pieces of factual information. Similarly, Tetlock (1983, 1993) uses the term *integrative complexity* to describe the degree to which considerations related to an issue are interconnected. Luskin (1987) also defines political sophistication based on the structure of individual belief systems,

¹A similar approach is taken by Colombo (2016) who investigates the competence of Swiss citizens voting in policy referenda. Colombo conceptualizes competence as a voter's ability to justify his or her political decisions, and measures the concept by manually coding open-ended responses to survey questions.

arguing that they can vary on three separate dimensions: (1) their *size* – i.e. the number of cognitions, (2) their *range* – i.e. the dispersion of cognition over categories, and (3) their *constraint* – i.e. the extent to which cognitions are interconnected in a meaningful way. Political sophistication, in turn, is seen as the conjunction of these dimensions: “A person is politically sophisticated to the extent to which his or her [political belief system] is large, wide-ranging, and highly constrained.” ([Luskin, 1987](#), 860).

Overall, this body of work suggests that differences in sophistication should be reflected in the way individuals describe and justify their political beliefs. Crucially, a measure of sophistication that is based on how individuals discuss their preferences in their own words can be directly applied in various settings to target specific political tasks such as choosing between candidates, parties, or policy propositions. Rather than having to devise a new set of questions that attempt to capture information necessary to make competent decisions, we can simply analyze how respondents elaborate on their related preferences in verbatim.

2 Measuring Discursive Sophistication

How would a politically sophisticated person who engages in in-depth processing discuss his or her views compared to a less informed individual? Consider a survey where respondents are asked to describe their attitudes toward specific policies or candidates running for office in a set of open-ended items. In such a scenario, the structure of individual political belief systems (i.e., size, range, and constraint) as well as the level of motivation to engage in more elaborate reasoning should be reflected in their verbatim responses. In the following, I discuss three different attributes of open-ended survey responses that should be indicative of sophistication in attitude expression.

First of all, sophisticated individuals should be able to elaborate more on their political attitudes. If people possess a large, wide-ranging, and constrained belief system, they should be able to recall a large number of *considerations* related to political actors or issues. I rely on the structural topic model framework ([Roberts et al., 2014](#)) to extract the number of topics

mentioned by each respondent in a survey.² First, denote \mathcal{W}_i as the set of words contained in a response of individual i . Each word $w \in \mathcal{W}_i$ is assigned to a topic $t^* \in \{1, \dots, T\}$, such that $P(t^*|w, X_i) > P(t|w, X_i) \forall t \neq t^*$.³ In other words, each unique term in a response is assigned to the topic that has the highest likelihood of having generated that term, given the model. The set of topics that are mentioned by respondent i across all words in \mathcal{W}_i can then be denoted as \mathcal{T}_i^* and the number of considerations can be written as:

$$\text{considerations}_i = \frac{|\mathcal{T}_i^*|}{\max_i |\mathcal{T}_i^*|}. \quad (1)$$

The measure is re-scaled to range from zero to one by dividing raw count of topics by the maximum number of topics observed across individuals.

However, sophisticated respondents should not only be able to mention a larger number of raw considerations when discussing politics. The level of sophistication should also be reflected in the *word choice* describing the underlying issues. Individuals who possess a constrained system of beliefs should be more inclined to use terms that are highly descriptive of a given topic (e.g., the *economy* or *taxes*) rather than broad terms that could be attributed to any topic and are not clearly related to politics. Highly descriptive word choice is conceptualized as the sum of term likelihoods $P(w|t^*)$ given topic assignments over the entire set of words in \mathcal{W}_i :

$$\text{word choice}_i = \frac{\sum_{\mathcal{W}_i} P(w|t^*)}{\max_i [\sum_{\mathcal{W}_i} P(w|t^*)]} \quad (2)$$

Again, the measure is re-scaled to range from zero to one by dividing all values by the empirical maximum observed across all individuals in the data.

Lastly, sophisticated individuals should hold opinions about each political actor or policy that they are asked to discuss. Given a set of multiple open-ended probes focusing on different issues,

²See below for more information on the set of open-ended responses, pre-processing choices, as well as on the topic model specification.

³Note that $P(t|w, X_i) = \frac{P(w|t)P(t|X_i)}{P(w|X_i)}$. In the context of structural topic models, X_i denotes the covariates used to predict individual topic prevalence (see [Roberts et al., 2014](#), for details).

sophisticates should be able to express their attitudes towards each question in terms of both approval or disapproval. Responses that reflect high levels of sophistication should therefore display a greater level of *opinionation*, which is conceptualized as the diversity of relative lengths for each open-ended response (specified as the Shannon entropy):

$$\text{opinionation}_i = \frac{-\sum_{j=1}^J p_{ij} \ln p_{ij}}{\ln J} \quad (3)$$

where p_{ij} is the proportion of words in the response of individual i to question $j \in \{1, \dots, J\}$ relative to the overall size of the individuals' response. The variable ranges from 0 (only one question was answered) to 1 (all questions were answered with the same word length per answer).

Together, the three measures form a composite metric of sophistication in political attitude expression by calculating their respective average for each respondent. Like each individual component, the resulting *discursive sophistication* score ranges from 0 to 1:

$$\text{discursive sophistication}_i = \frac{1}{3}(\text{considerations}_i + \text{word choice}_i + \text{opinionation}_i). \quad (4)$$

Overall, a highly sophisticated individual can be expected to respond to a set of open-ended items by giving a more elaborate response that focuses on multiple considerations using terms that are highly descriptive of each topic and addresses his or her attitudes towards all relevant political actors or policies more or less equally.⁴

⁴Note that this approach differs from recent work on sophistication in speeches and other sources of political texts (e.g., [Spirling, 2016](#); [Benoit, Munger, and Spirling, 2017](#)) as it explicitly targets complexity independent of pure linguistic style.

3 An Overview of Data Sources and Open Ended Items

The measure of discursive sophistication is validated using multiple surveys employing different sets of open-ended questions. Each survey focuses on sophistication in the context of distinct political tasks, namely the evaluation of (1) candidates running for public office, (2) broad issue areas such as health care and gun legislation, and (3) policy referenda. The data sets and items used to compute discursive sophistication are briefly described below.⁵

3.1 2012 & 2016 American National Election Study

The main analyses are based on the 2012 and 2016 wave of the American National Election Study (ANES), which consist of a representative survey of about 5000 adults in the months before the US Presidential election in each year. About 2000 respondents in both waves participated in face-to-face interviews while the remaining respondents filled out the survey online. For the purpose of the present analyses, I rely on the pooled datasets while controlling for differences in survey mode. The measure of discursive sophistication is based on open-ended questions in which respondents were asked in the pre-election wave of the survey to list anything in particular that they like/dislike about the Democratic/Republican party as well as anything that might make them vote/not vote for either of the Presidential candidates. They were probed by the interviewer asking “anything else?” until the respondent answered “no.” Overall, there are a total number of 8 open-ended responses where individuals described their beliefs and attitudes towards political actors. Individuals who did not respond to all of the open-ended items (420 in 2012; 204 in 2016), or who responded in Spanish (228 in 2012; 43 in 2016), are excluded from the analysis.

⁵See Appendix A for descriptive information on open-ended responses in each dataset, structural topic model results, and individual components of discursive sophistication. Appendix B contains further details on pre-processing steps and modeling choices for the structural topic models as well as robustness checks, which include preText analyses proposed by Denny and Spirling (2018). Lastly, Appendix C provides information on the remaining variables included in the analyses.

3.2 2015 YouGov Survey

In order to replicate and extend the main analyses, I rely on a separate nationally representative survey employing an alternative set of open-ended responses. The data was collected by YouGov in December 2015 and contains responses of 1000 U.S. citizens.⁶ As part of this study, respondents were asked to describe their attitudes towards two prominent political issues that were discussed frequently in the media. First, they were asked in a closed format whether they favor or oppose stricter gun laws. Subsequently, they were asked to respond to the following two questions:

- Still thinking about the question you just answered, what thoughts came to mind while you were answering that question? Please try to list everything that came to mind.
- Thinking about the mass shootings that have occurred in the U.S. in the last few years, what factors do you think are responsible for the shootings?

Second, the respondents reported on their attitudes towards the Affordable Care Act in a closed format and were then asked to elaborate in their own words by answering the following questions:

- Still thinking about the question you just answered, what thoughts came to mind while you were answering that question? Please try to list everything that came to mind.
- For decades, experts have observed that the United States spends far more per person on health care than any other country. However, the U.S. falls behind on most measures of health care outcomes, such as life expectancy. What factors do you think are responsible for the state of our health care system?

Here, discursive sophistication is computed based on the verbatim responses to the four preceding questions using the same procedures described above. Compared to the open-ended likes/dislikes items included in the 2012 and 2016 ANES, the questions directly address considerations related to specific policy issues that were prominent in the political discourse at the time of the survey. Respondents who did not provide an answer to any of the open-ended questions were removed from the analysis (48).

⁶See Clifford and Jerit (2018) for details on the study.

3.3 Swiss Referendum Survey

Lastly, I examine survey data on Swiss citizens justifying their vote choices on multiple referenda used in a recent analysis by Colombo (2016). The author compiled a data set of cross-sectional surveys administered in Switzerland after national popular votes on multiple policy propositions. The original surveys were conducted as representative samples after each of thirty-four national policy votes that were held between 2008 and 2012 resulting in a total of about 27,000 observations. However, respondents were only asked to justify their decision for or against a given proposition in verbatim if they participated in the vote in the first place. As such, about 5,000 individuals in the data set did not provide an open-ended response. The remaining respondents were asked to describe the main reason as well as additional justifications for their decision in two separate items. As before, discursive sophistication is computed based on verbatim responses to both questions.

4 A First Look at Discursive Sophistication

Before turning to the validation, I begin by directly comparing discursive sophistication to alternative metrics of political knowledge in the 2012 and 2016 ANES. The standard approach to measuring political knowledge in surveys is to ask a set of factual questions about political institutions. The ANES surveys include such a basic item battery, inquiring for example about the number of times an individual can be elected President of the United States, or how the current U.S. federal budget deficit compares to the deficit in the 1990s. I combine responses on these items to form an additive index of *factual knowledge* about politics. As an additional benchmark, I consider *interviewer assessments* of each respondent's political sophistication (c.f., Bartels 2005 for an example of a study that relies on interviewer assessments; but see also Ryan 2011).⁷

Figure 1 compares discursive sophistication to the conventional knowledge metrics for both surveys. Each figure presents scatterplots between individual measures (lower triangular), univari-

⁷Interviewer assessments were only recorded in the face-to-face sample of the ANES.

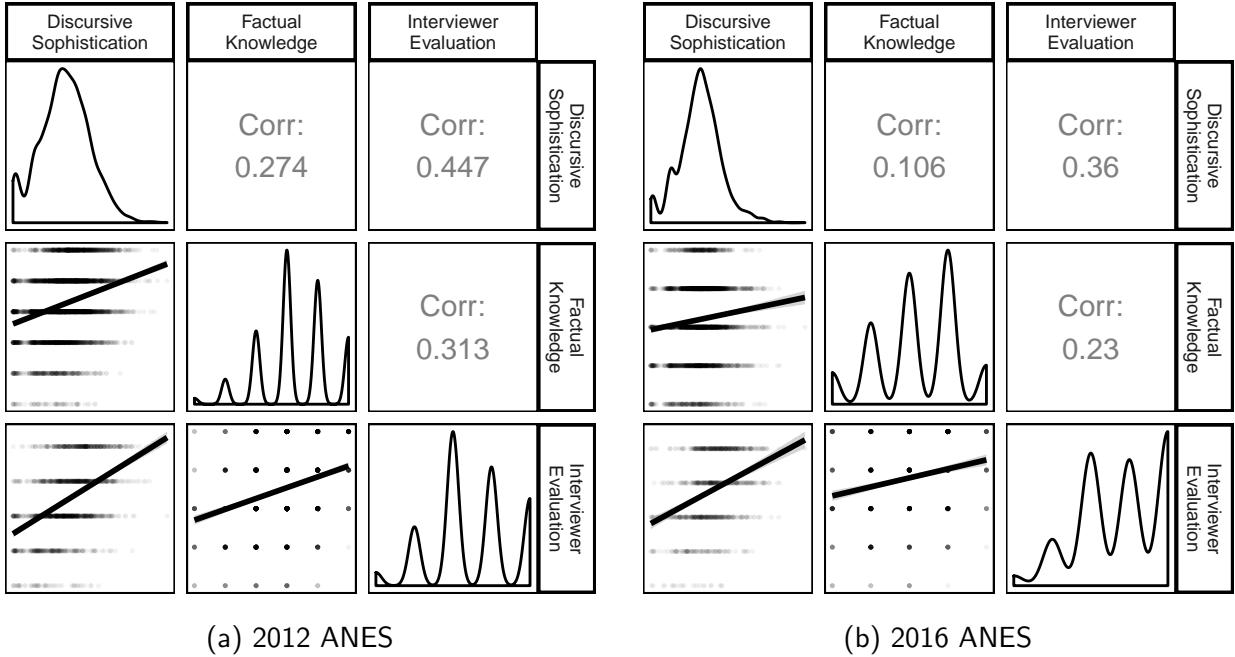


Figure 1: Correlation matrix of conventional political knowledge metrics and discursive sophistication. The plots on the diagonal display univariate densities for each variable. The panels in the lower triangular display the scatter plot of two measures as well as a linear fit. The upper triangular displays the correlation coefficient. All correlations reported are statistically significant with $p < .05$.

ate densities (diagonal), and correlation coefficients (upper triangular). The measure of discursive sophistication is positively correlated with both conventional metrics while capturing some additional variation. Interestingly, there is a stronger correlation between discursive sophistication and interviewer evaluations than between factual knowledge and interviewer evaluations ($r = .45$ vs. $r = .31$ in 2012, and $r = .36$ vs. $r = .23$ in 2016). The open-ended measure therefore captures characteristics that influence subjective assessments of sophistication. Interviewers certainly form their impressions throughout the entire survey, but a respondent's verbatim answers seems to be more influential for subsequent knowledge assessments than a respondent's performance on the factual knowledge questions.

Overall, while discursive sophistication and the alternative measures are clearly correlated, the relationship between each metric is far from perfect. To provide some intuition as to whether the variation in discursive sophistication is theoretically meaningful, I present an example of

open-ended responses of two individuals in the 2016 ANES who identified as Republicans and scored equally on the factual knowledge score (3 out of 4 correct responses), but varied highly in discursive sophistication. The results are presented in Table 1.

	A: Low Sophistication Response	B: High Sophistication Response
Clinton (+)		Politician.
Clinton (-)	The fact that she has links to Al-Qaeda.	Caught in lies.
Trump (+)		Says what he thinks.
Trump (-)	He is going to start a civil war. I feel like he is racist.	Reality TV star, poor businessman
Democrats (+)		Middle class minded.
Democrats (-)		Too many handouts.
Republicans (+)		Economic growth conscious.
Republicans (-)		For the big business.
Disc. Soph.	0.162	0.461

Table 1: Example of open-ended responses for low and high scores on discursive sophistication with equal factual knowledge scores (3 out of 4 correct responses). Column A displays the verbatim responses of an individual who scored low on discursive sophistication and column B displays the verbatim responses of an individual who scored high on the open-ended measure. Each row represents one of the likes/dislikes items included in the analysis. Note that the responses in this table were slightly redacted for readability (spelling errors removed, etc.).

Each row in the table represents one of the open-ended responses (like/dislike for each candidate/party). Column A displays the responses of an individual who scored low on discursive sophistication and column B displays the responses of a high scoring individual. Cells are empty if a respondent refused to provide a response. Even though both individuals are measured to have equal factual political knowledge, there are systematic differences in their response behavior that can be attributed to their political sophistication. Overall, respondent A provided a less elaborate response, only focused on a narrow range of issues, and only reported attitudes on two items. Irrespective of whether one agrees with the specific statements or whether they are factually accurate (e.g., Clinton's connection to Al-Qaeda), A's response pattern is suggestive of a less sophisticated political belief system and a lower level of motivation to engage in in-depth processing about both parties and candidates. Overall, this initial result suggests that the variation in discursive sophistication captures meaningful differences in response behavior that overlaps with traditional knowledge metrics while displaying some unique variation. The following sections will

show that this variation is also politically consequential.

5 Discursive Sophistication and Political Competence

I validate the measure of discursive sophistication by directly examining its effects on individual competences to perform political tasks in modern democracies (c.f., Lupia, 2006, 2015). More specifically, I consider the potential role of political sophistication in promoting (1) engagement and participation in politics, (2) the ability to incorporate new information, (3) precise positioning of candidates running for election, and (4) well-justified policy preferences. In the following, each point will be addressed using one of the three data sets described above.

5.1 Engagement and Participation in Politics

Political sophistication is often argued to promote individual engagement and participation in politics. In fact, factual knowledge items have been validated in the past based on their strong relationship with outcomes such as turnout and other forms of participation (c.f., Lupia, 2015, 230–233). Figure 2 compares the effects of discursive sophistication and factual knowledge in the 2012 and 2016 ANES on four dependent variables related to political engagement: turnout, non-conventional participation, internal efficacy, and external efficacy. The model predicting turnout is estimated via logistic regression while the estimates for the three remaining dependent variables are based on OLS. Each model equation includes both sophistication measures while controlling for gender, education, income, age, race, church attendance, survey mode (face-to-face vs. online), as well as the Wordsum vocabulary score measuring verbal intelligence.

Each panel displays the expected difference in the respective dependent variable for a two standard deviation increase in each sophistication measure, while holding all other variables constant at their means. Overall, discursive sophistication is a stronger predictor of turnout, non-conventional participation, as well as (to a lesser extent) internal and external efficacy. In the 2012 ANES, the positive effect of factual knowledge on participation is statistically indistinguish-

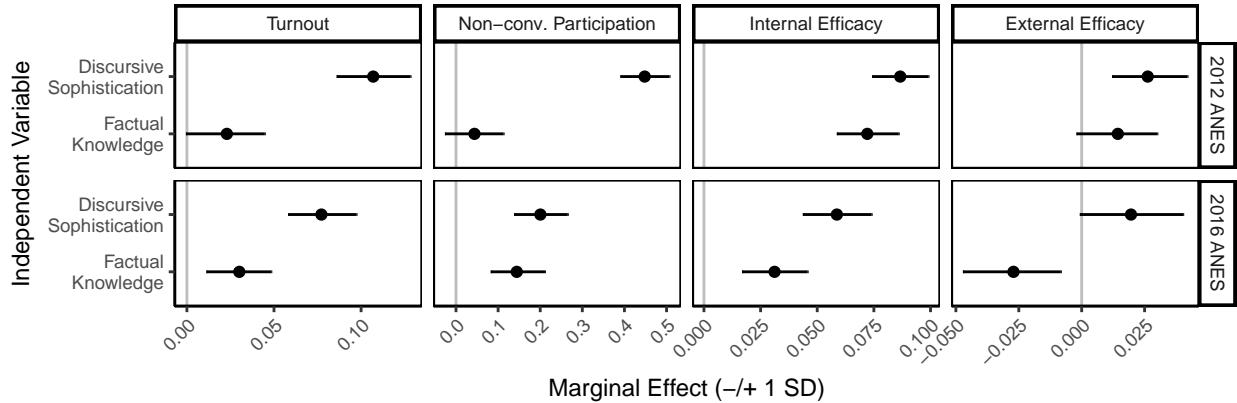


Figure 2: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy in the 2012 and 2016 ANES. For each dependent variable, the figure displays the change in expected values after increasing each sophistication measure from -1 to +1 standard deviation from its mean (including 95% confidence intervals). Model estimates are based on logistic regression (turnout) or OLS (non-conventional participation, internal efficacy, external efficacy). Both sophistication measures are included simultaneously while controlling for gender, education, income, age, race, church attendance, survey mode, and Wordsum vocabulary scores. Full model results are displayed in the appendix, Tables 2 and 3.

able from zero when controlling for discursive sophistication. Furthermore, there is a negative effect of factual knowledge on external efficacy in the 2016 ANES. In contrast, the positive effect of discursive sophistication on external efficacy is more consistent with previous research. Considering these initial results, a potential concern may be that discursive sophistication is confounded by personality characteristics that influence verbatim response patterns as well as engagement. Appendix D provides additional analyses controlling for such factors that might drive verbosity (extraversion and being reserved) as well as individual response length itself. The substantive conclusions remain unchanged.

5.2 Incorporation of New Information

Competent citizens should not only engage in politics but are also expected to be sufficiently informed about the issues of the day. As such, they have to be attentive to their media environments and incorporate potentially relevant new information about parties, office-holders, and policies. Indeed, Zaller (1990, 1992) and others argue that tests of factual information about

politics are the best available proxy for awareness. In this analysis I draw on the 2015 YouGov study to explore whether discursive sophistication or factual knowledge serves as a better predictor of people's ability to incorporate new information from media sources. As part of the survey, respondents were asked to read a newspaper article about a fictional infectious disease and were subsequently asked to answer questions about information provided in the article (e.g. regarding symptoms, modes of contraction etc.). I compute an additive index counting the pieces of information that were correctly recalled (*information retrieval*) as a measure of the ability to retrieve information from a news article on a non-partisan issue that is related to public health policies.

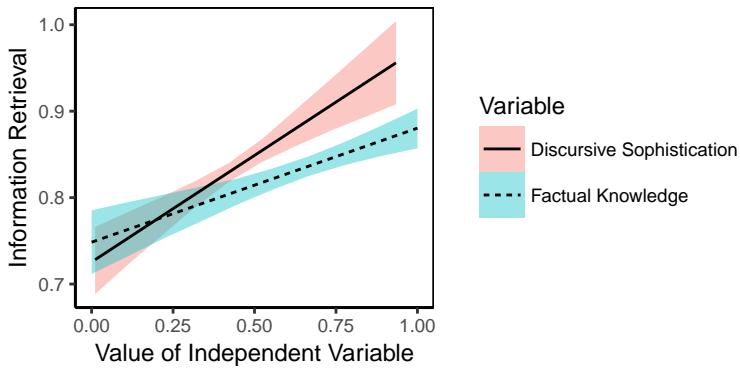


Figure 3: Expected information retrieval in the 2015 YouGov Study as a function of political sophistication (including 95% confidence intervals). Estimates are based on a linear regression model controlling for education, income, age, church attendance, gender, and race. Full model results are displayed in the appendix, Table 4.

Figure 3 displays the relationship between political sophistication and disease information retrieval in the 2015 YouGov study. Estimates are based on a linear regression model controlling for education, income, age, church attendance, gender, and race. As a benchmark for discursive sophistication, I again consider the effect of factual knowledge based on a battery of eight items similar to the knowledge questions in the ANES. Both discursive sophistication as well as factual knowledge increase the amount of information individuals are able to recall from a news article discussing a fictional disease. Similar to the previous results, the effects are stronger for discursive sophistication than for factual knowledge scores. The degree to which citizens discuss their own political beliefs in a more elaborate manner is not only a stronger predictor of political engagement

but also serves as a better proxy for the ability to incorporate new information about a non-partisan issue.

5.3 Precise Positioning of Candidates

Citizens who are attentive to politics and able to incorporate new information should ultimately be better informed about the policies put forward by parties and political elites. This is a crucial component of citizen competence in representative democracies since precise knowledge about the policy positions held by candidates who are running for office allows voters to hold them accountable. Figure 4 presents the results of multiple heteroskedastic regressions where the error variance in candidate placements on multiple issues included in both ANES waves (general ideology, government spending, defense spending, health insurance policy, job guarantee, government assistance to Blacks, environment vs. jobs trade-off) is modeled as a function of discursive sophistication as well as factual knowledge (see [Jacoby, 2006](#), for a similar procedure). More formally, each model for a given candidate placement on a specific policy issue takes the following form:

$$y \sim N(\mu, \sigma) \quad (5)$$

$$\mu = X\beta \quad (6)$$

$$\log(\sigma) = Z\gamma, \quad (7)$$

where y is the vector of policy placements across respondents, X is a matrix of covariates predicting average candidate placements μ (self-placement, education, income, age, church attendance, gender, race, and survey mode), Z denotes the covariates predicting the error variances σ (discursive sophistication, factual knowledge, Wordsum score), and β and γ are the parameters to be estimated.

The figure displays the estimated reduction in error variances of candidate placements when each sophistication measure is increased by two standard deviations. Larger negative values indi-

cate a stronger reduction in error variances and hence more precise candidate placements. Both factual knowledge and discursive sophistication significantly decrease error variances in policy placements of presidential candidates. Some interesting differences, however, emerge when comparing both waves of the ANES. In the 2012 election, discursive sophistication in open-ended responses was a slightly weaker predictor of precise candidate placements than performance on factual knowledge quizzes across multiple issues. This picture is reversed in the 2016 election, where more elaborate open-ended responses were strongly predictive of precise candidate placements. Factual knowledge, on the other hand, did not significantly improve the precision of candidate placements for multiple issues.

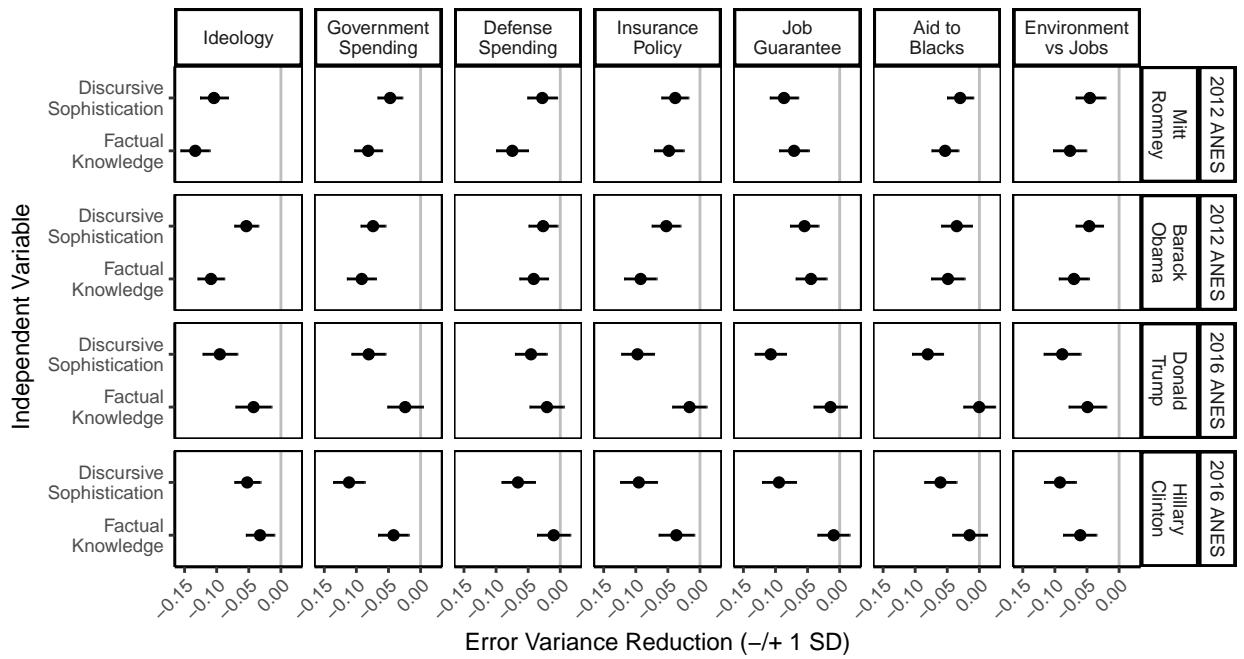


Figure 4: Error variance reduction in candidate placements on multiple issues in the 2012 and 2016 ANES. The figure displays the difference in estimated error variances after increasing each sophistication measure from -1 to $+1$ standard deviation from its mean (including 95% credible intervals). Models are estimated in Stan using non-informative priors. Detailed model results including convergence statistics are displayed in the appendix, Tables 5 and 6.

5.4 Well-Justified Policy Preferences

Beyond keeping track of the candidates' positions, competent citizens should be knowledgeable about the underlying policies themselves and be able to justify their own preferences. Here, I explore the extent to which high levels of discursive sophistication correspond to well-justified policy preferences in open-ended responses. As mentioned above, the Swiss surveys included items that asked respondents to explain why they voted in favor or against a given proposition in multiple policy referenda. To corroborate the face validity of discursive sophistication, I examine whether the measure is related to Colombo's (2016) manual coding of the respondents' *level of justification*, which assessed the content, elaboration, and complexity of open-ended responses.

The results are presented in Figure 5. Since the Swiss post-referendum surveys were conducted in three different languages (German, French, and Italian), I computed the measure of discursive sophistication separately for each group of respondents. The figure displays the distribution of discursive sophistication for each level of justification coded by Colombo (2016) as well as the correlation coefficients for both respective variables. Across all three language groups, discursive sophistication is systematically higher among respondents with the highest level of justification and both measures are positively correlated ($r = 0.29, 0.25$, and 0.35 , respectively). The proposed measure of discursive sophistication therefore shows a high degree of correspondence with individual levels of justification assessed by independent manual coders.

To summarize, the results presented thus far indicate that discursive sophistication shares common characteristics with factual political knowledge measures. Compared to conventional metrics, the proposed measure performs at least as well as a predictor of essential competences that allow citizens to engage successfully in politics. In fact, discursive sophistication is a stronger predictor of certain outcomes (such as different forms of political participation) than conventional knowledge scores. In the following, I turn to an application to illustrate how discursive sophistication can help refine important previous insights from the literature on political knowledge.

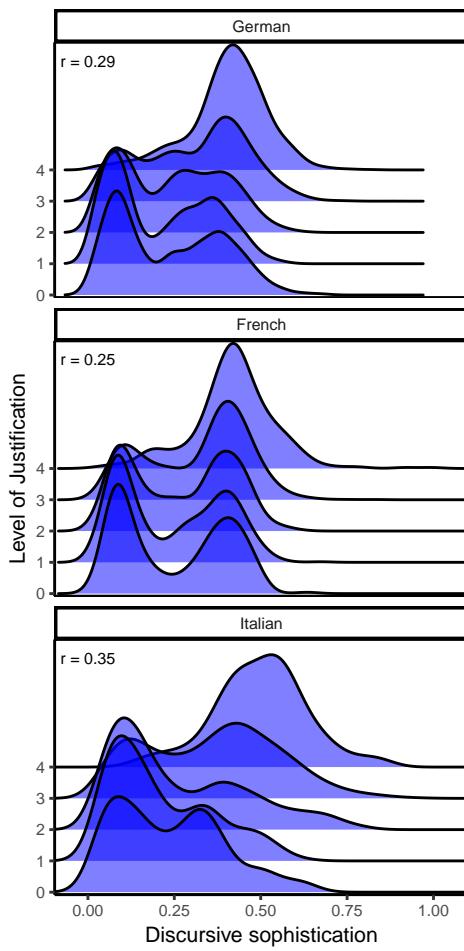


Figure 5: Discursive sophistication and manually coded level of justification ([Colombo, 2016](#)) in Swiss post-referendum surveys. The plot compares kernel densities of discursive sophistication for each manually coded level of justification.

6 Application: The Gender Gap in Political Knowledge

A common finding in public opinion research is the fact that women have lower levels of observed political knowledge than men. For example, Verba, Burns, and Schlozman (1997) report that women score lower on political information, interest, and efficacy, which decreases their respective levels of political participation. Since gender differences in political information and interest can only partly be explained by resource-related factors such as individual levels of education, the authors diagnose a “genuine difference in the taste for politics” between men and women, which they suspect to be driven largely by socialization (see also Wolak and McDevitt, 2011). Indeed, Dow (2009, 117) describes the systematic gender differences in knowledge “one of the most robust findings in the study of political behavior.”

The discussion revolving around this apparent gender gap is closely intertwined with the methodological debate about measuring political knowledge. For example, Mondak and Anderson (2004) suggest that women are more likely to report that they do not know the answer to a recall question whereas men are more inclined to guess. Correcting for the systematic differences in the propensity to guess, however, mitigates the gender gap in knowledge but does not eliminate it completely (see also Lizotte and Sidman, 2009). Other aspects of the survey context have been shown to affect gender differences in political knowledge. For example, McGlone, Aronson, and Kobrynowicz (2006) present evidence that the gender gap is exacerbated in an environment that induces stereotype threat, for example if women are aware of the fact that the study focuses on gender differences or if they are interviewed by a male interviewer. However, gender differences are not only induced by *how* researchers ask their questions, but also by the question *content* itself. For example, Dolan (2011) argues that the gap can be closed by focusing on gender-relevant political knowledge items such as information about women’s representation in the federal government (see also Gruber, 2001; Fraile, 2014; Jerit and Barabas, 2017). Similarly, Stolle and Gidengil (2010) report that the gender gap disappears when people are asked about more practical issues related to the government (e.g., benefits and services).

Overall, the gender gap has been shown to be influenced by how we ask for political information

in surveys, as well as the kind of knowledge that is required for a correct response. Indeed, a comprehensive cross-national analysis of election studies in 47 countries between 1996 and 2011 suggests that question format and content account for large portions of the variance of gender disparities in political knowledge ([Fortin-Rittberger, 2016](#)).

6.1 Descriptive Results

How do men and women compare on the different metrics of political sophistication in the surveys analyzed in the present study? Figure 6 displays the average levels of discursive sophistication as well as conventional metrics comparing both genders. While we observe a sizable and statistically significant gender gap for factual knowledge in both ANES surveys, this difference disappears for discursive sophistication. These results are replicated in the 2015 YouGov survey. As before, we observe a significant gender gap in factual knowledge which disappears using the discursive measure. Of course, it is important to ask whether this absence of a gender gap in discursive sophistication is theoretically meaningful or rather an artifact of the measurement approach itself.

One way to investigate this question is to explore gender differences in discursive sophistication using the [Colombo \(2016\)](#) data and comparing them to her manually coded measure. That way, we can not only examine whether the lack of a gender gap in discursive sophistication replicates using an additional survey, but also check whether there is an equivalent lack of gender differences in Colombo's alternative measure of citizen competence in direct democracies. If discursive sophistication captures a person's motivation to undertake in-depth processing and form quality opinions (and assuming these characteristics do not differ by gender), there should be no difference between men and women on either metric (discursive sophistication and Colombo's measure). As shown in the bottom row of Figure 6 there are indeed no significant gender differences on *both* metrics across all three languages in the Swiss referendum surveys. The absence of a gender gap is consistent whether open-ended responses are coded manually or using the proposed measure of discursive sophistication.

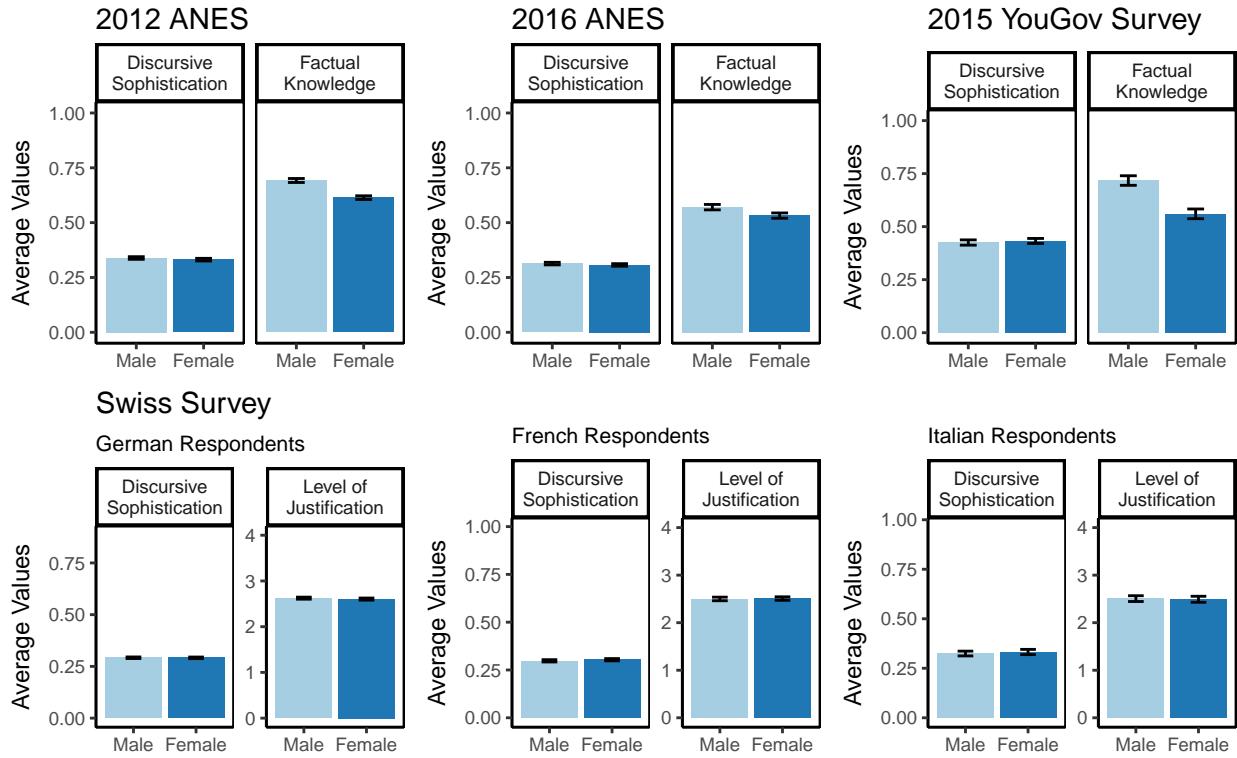


Figure 6: The gender gap in political sophistication. The figures display mean levels of sophistication for each measure comparing men and women (including 95% confidence intervals). Gender differences in factual knowledge in the 2012/2016 ANES and 2015 YouGov survey (top row) are statistically significant with $p < .05$. Gender differences in discursive sophistication and manually coded levels of justification (Colombo, 2016) are not statistically significant.

6.2 Controlling for Alternative Explanations

Prior research suggests that at least part of the gender gap in political knowledge can be attributed to real discrepancies in resources and engagement. To the extent that differences between men and women can be explained by these underlying factors, they are less likely to be an artifact of the measurement of knowledge itself. Accordingly, we need to control for determinants of political knowledge to provide a more comprehensive examination of the veracity of observed gender differences. Figure 7 displays estimated effects of various potential common determinants of factual knowledge and discursive sophistication on both measures. Previous studies consistently showed that political information levels are positively related to high media exposure, frequent political discussions, education, and income. Furthermore, I include age, race, church attendance,

and survey mode (face-to-face vs. online) as additional control variables.

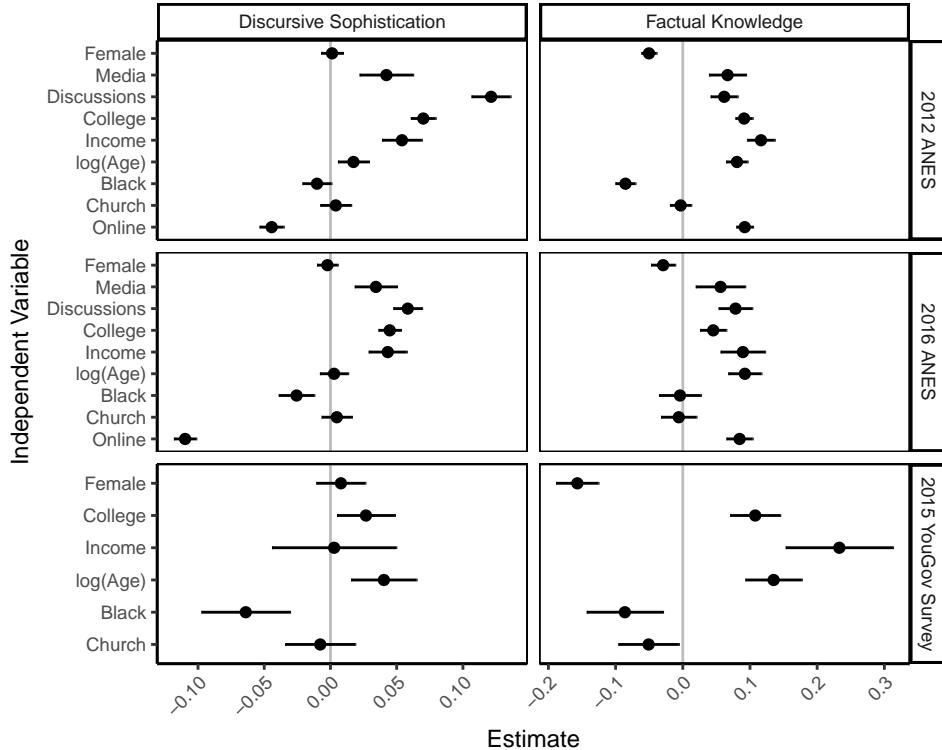


Figure 7: Common determinants of political sophistication. Estimates are OLS regression coefficients with 95% confidence intervals. Dependent variables are discursive sophistication as well as conventional metrics of political knowledge. Full model results are displayed in the appendix, Tables 7 and 8.

After controlling for common determinants, discursive sophistication again reveals no significant differences between men and women in both ANES surveys as well as the 2015 YouGov study. The gender gap in factual political knowledge, however, persists and is substantively as well as statistically significant after controlling for various resource-related factors. Even though women do not perform as well as men on political quizzes, they do not differ substantially in complexity and sophistication when they describe their political preferences. The effects of the remaining variables are quite similar across both measures and different surveys. Knowledge and sophistication is significantly higher among respondents who are more exposed to political news media, discuss politics frequently, are more educated, and have higher income.⁸ Overall, the

⁸An interesting deviation, however, is the effect of survey mode in the 2012 and 2016 ANES. Respondents in online surveys score significantly higher on factual knowledge than in face-to-face interviews. This difference

finding that determinants of political sophistication are consistent across models lends additional validity to the open-ended measure.

To summarize, we only observe a significant gender gap when looking at conventional recall-based measures, a result that previous research (at least partly) attributed to the content (i.e., focusing on issues that are less relevant to women) and format (i.e., stereotype-threat and guessing) of the question batteries. When using the alternative measure—discursive sophistication—any evidence for systematic differences between men and women disappears.

6.3 Explaining the (Lack of the) Gender Gap

If it is the case that women are able to close the gender gap in discursive sophistication because they are able to focus on different considerations that are salient to them when discussing their political preferences, we should observe systematic variation in the issues men and women discuss in open-ended responses. Based on the structural topic model used to compute discursive sophistication, I now examine the subset of topics that showed the largest absolute gender difference in topic prevalence in the 2012 and 2016 ANES. The results are displayed in Figure 8.

Positive coefficients indicate that women are more likely than men to mention a given topic, and vice versa. As such, the top six topics are more prevalent among men and the bottom six have a higher probability to be mentioned by women. Each coefficient is labeled with the five highest probability terms related to the topic to illustrate its content. Across both ANES studies, women were less likely than men to discuss foreign affairs, economic issues, or the Supreme Court. Instead, they focused on issues related to women's rights, equality, or health care. The considerations taken into account by women when discussing their political preferences are therefore clearly different from men's and—crucially—the issues raised by men happen to be more aligned with what political scientists often deem as necessary information (i.e., pertaining to

can be attributed to the fact that individuals are able to look up answers for factual knowledge questions while taking an online survey (c.f., Clifford and Jerit, 2016). For discursive sophistication, on the other hand, individuals perform better in the face-to-face survey. Open-ended answers in online surveys may be less elaborate because respondents have to manually type their responses.

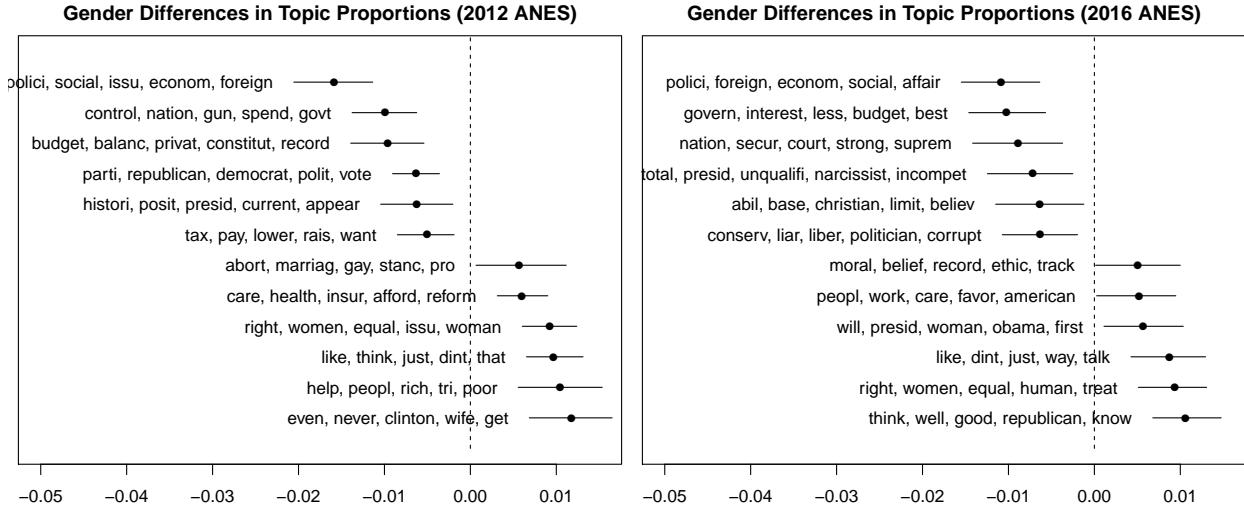


Figure 8: Gender differences in topic proportions in open-ended responses based on the structural topic model used to compute discursive sophistication (including 95% confidence intervals). Coefficients indicate the difference in predicted topic prevalence among men and women; positive values indicate higher prevalence among women. Labels are based on the five highest probability terms related to the topic.

the economy, institutions, elites, etc.). Yet, from a normative perspective, there is no reason to assume that one set of issues should be more important for citizens when forming their political preferences and making competent voting decisions.

7 Conclusion

Political scientists should worry less about pure levels of *factual knowledge* and instead focus on the necessary conditions for individuals to make *competent* decisions. Competence in the context of political decision-making and voting requires citizens to hold informed attitudes about their representatives. Factual knowledge about political institutions might be a useful proxy for competence in certain scenarios. However, it cannot address directly whether individuals hold well-considered opinions about political actors they try to hold accountable. In comparison, the measure of discursive sophistication proposed here is agnostic about the specific contents of people’s beliefs, but directly captures the complexity of individual attitude expression. Furthermore, it can be easily applied to assess sophistication in any decision-making context (such as a policy

referendum or a local election) by fielding targeted open-ended questions related to the relevant underlying beliefs and preferences.

The findings presented in this paper show that conventional knowledge indices and the discursive measure share a substantial amount of variance. However, they are far from being identical and capture different aspects of sophistication. Most importantly, using the discursive measure, evidence for the gender gap commonly reported using factual knowledge scales disappears. Women might know fewer facts about political institutions, but they do not differ substantively in the complexity of their expressed political beliefs. The fact that women perform just as well as men on discursive sophistication across various surveys can be attributed to the fact that they focus on different considerations when evaluating political parties and candidates. This issue has long been recognized in the literature (e.g., [Graber, 2001](#); [Dolan, 2011](#)), but it cannot be properly addressed while relying exclusively on off-the-shelf political knowledge batteries. As discussed at the outset, [Zaller \(1992\)](#) and others made the argument that testing for factual information provides the best measure of political awareness as it captures “what has actually gotten into people’s minds, which, in turn, is critical for intellectual engagement with politics” (21). The results presented in this paper suggest that a direct examination of open-ended responses provides a viable alternative approach.

References

- Barabas, Jason, Jennifer Jerit, William Pollock, and Carlisle Rainey. 2014. "The Question(s) of Political Knowledge." *American Political Science Review* 108 (04): 840–855.
- Bartels, Larry M. 2005. "Homer gets a tax cut: Inequality and public policy in the American mind." *Perspectives on Politics* 3 (1): 15–31.
- Benoit, Kenneth, Kevin Munger, and Arthur Spirling. 2017. "Measuring and Explaining Political Sophistication Through Textual Complexity." Available at SSRN: <https://ssrn.com/abstract=3062061> or <http://dx.doi.org/10.2139/ssrn.3062061>.
- Bolsen, Toby, James N. Druckman, and Fay Lomax Cook. 2014a. "The Influence of Partisan Motivated Reasoning on Public Opinion." *Political Behavior* 36 (2): 235–262.
- Cappella, Joseph N, Vincent Price, and Lilach Nir. 2002. "Argument repertoire as a reliable and valid measure of opinion quality: Electronic dialogue during campaign 2000." *Political Communication* 19 (1): 73–93.
- Clifford, Scott, and Jennifer Jerit. 2016. "Cheating on Political Knowledge Questions in Online Surveys: An Assessment of the Problem and Solutions." *Public Opinion Quarterly*.
- Clifford, Scott, and Jennifer Jerit. 2018. "Disgust, Anxiety, and Political Learning in the Face of Threat." *American Journal of Political Science* (forthcoming): <https://doi.org/10.1111/ajps.12350>.
- Colombo, Céline. 2016. "Justifications and Citizen Competence in Direct Democracy: A Multi-level Analysis." *British Journal of Political Science*: 1–20.
- Converse, Philip E. 1964. "The Nature of Belief Systems in Mass Publics." In *Ideology and Discontent*, ed. David E. Apter. New York: Free Press.
- Cramer, Katherine J, and Benjamin Toff. 2017. "The Fact of Experience: Rethinking Political Knowledge and Civic Competence." *Perspectives on Politics* 15 (3): 754–770.
- Delli Carpini, Michael X., and Scott Keeter. 1996. *What Americans Know about Politics and Why It Matters*. New Haven, London: Yale University Press.
- Denny, Matthew J., and Arthur Spirling. 2018. "Text Preprocessing For Unsupervised Learning: Why It Matters, When It Misleads, And What To Do About It." *Political Analysis* 26 (2): 168–189.
- Dolan, Kathleen. 2011. "Do women and men know different things? Measuring gender differences in political knowledge." *The Journal of Politics* 73 (01): 97–107.
- Dow, Jay K. 2009. "Gender differences in political knowledge: Distinguishing characteristics-based and returns-based differences." *Political Behavior* 31 (1): 117–136.
- Druckman, James N. 2014. "Pathologies of studying public opinion, political communication, and democratic responsiveness." *Political Communication* 31 (3): 467–492.

- Fortin-Rittberger, Jessica. 2016. "Cross-National Gender Gaps in Political Knowledge How Much Is Due to Context?" *Political Research Quarterly* 69 (3): 391–402.
- Fraile, Marta. 2014. "Does deliberation contribute to decreasing the gender gap in knowledge?" *European Union Politics* 15 (3): 372–388.
- Gilens, Martin. 2001. "Political ignorance and collective policy preferences." *American Political Science Review* 95 (02): 379–396.
- Graber, Doris A. 2001. *Processing politics: Learning from television in the Internet age*. University of Chicago Press.
- Jacoby, William G. 2006. "Value choices and American public opinion." *American Journal of Political Science* 50 (3): 706–723.
- Jerit, Jennifer, and Jason Barabas. 2017. "Revisiting the Gender Gap in Political Knowledge." *Political Behavior* 39 (4): 817–838.
- Kunda, Ziva, and Lisa Sinclair. 1999. "Motivated reasoning with stereotypes: Activation, application, and inhibition." *Psychological Inquiry* 10 (1): 12–22.
- Lee, Moontae, and David Mimno. 2014. "Low-dimensional embeddings for interpretable anchor-based topic inference." In *Proceedings of Empirical Methods in Natural Language Processing*. Citeseer.
- Lizotte, Mary-Kate, and Andrew H Sidman. 2009. "Explaining the gender gap in political knowledge." *Politics & Gender* 5 (02): 127–151.
- Lupia, Arthur. 2006. "How elitism undermines the study of voter competence." *Critical Review* 18 (1-3): 217–232.
- Lupia, Arthur. 2015. *Uninformed: Why people seem to know so little about politics and what we can do about it*. Oxford University Press.
- Luskin, Robert C. 1987. "Measuring political sophistication." *American Journal of Political Science* 31 (4): 856–899.
- Manning, Christopher D., Prabhakar Raghavan, Hinrich Schütze et al. 2008. *Introduction to Information Retrieval*. Cambridge: Cambridge University Press.
- McGlone, Matthew S, Joshua Aronson, and Diane Kobrynowicz. 2006. "Stereotype threat and the gender gap in political knowledge." *Psychology of Women Quarterly* 30 (4): 392–398.
- Mondak, Jeffery J, and Mary R Anderson. 2004. "The knowledge gap: A reexamination of gender-based differences in political knowledge." *Journal of Politics* 66 (2): 492–512.
- Mondak, Jeffrey J. 2001. "Developing valid knowledge scales." *American Journal of Political Science* 45 (1): 224–238.

- Pietryka, Matthew T, and Randall C MacIntosh. 2013. "An Analysis of ANES Items and Their Use in the Construction of Political Knowledge Scales." *Political Analysis* 21 (4): 407–429.
- Redlawsk, David P. 2002. "Hot cognition or cool consideration? Testing the effects of motivated reasoning on political decision making." *The Journal of Politics* 64 (04): 1021–1044.
- Roberts, Margaret E, Brandon M Stewart, and Dustin Tingley. 2014. "stm: R package for structural topic models." *Journal of Statistical Software* 1: 1–49.
- Roberts, Margaret E., Brandon M. Stewart, Dustin Tingley, Christopher Lucas, Jetson Leder-Luis, Shana Kushner Gadarian, Bethany Albertson, and David G. Rand. 2014. "Structural Topic Models for Open-Ended Survey Responses." *American Journal of Political Science* 58 (4): 1064–1082.
- Rosenberg, Shawn W. 1988. "The structure of political thinking." *American Journal of Political Science*: 539–566.
- Rosenberg, Shawn W, Dana Ward, and Stephen Chilton. 1988. *Political reasoning and cognition: A Piagetian view*. Duke University Press.
- Ryan, John Barry. 2011. "Accuracy and bias in perceptions of political knowledge." *Political Behavior* 33 (2): 335–356.
- Spirling, Arthur. 2016. "Democratization and Linguistic Complexity The Effect of Franchise Extension on Parliamentary Discourse, 1832–1915." *The Journal of Politics* 78 (1): 120–136.
- Stolle, Dietlind, and Elisabeth Gidengil. 2010. "What do women really know? A gendered analysis of varieties of political knowledge." *Perspectives on Politics* 8 (01): 93–109.
- Sturgis, Patrick, Nick Allum, and Patten Smith. 2008. "An experiment on the measurement of political knowledge in surveys." *Public Opinion Quarterly* 72 (1): 90–102.
- Tetlock, Philip E. 1983. "Cognitive style and political ideology." *Journal of Personality and Social Psychology* 45 (1): 118.
- Tetlock, Philip E. 1993. "Cognitive structural analysis of political rhetoric: Methodological and theoretical issues." In *Explorations in political psychology*. Duke University Press Durham.
- Verba, Sidney, Nancy Burns, and Kay Lehman Schlozman. 1997. "Knowing and caring about politics: Gender and political engagement." *The Journal of Politics* 59 (04): 1051–1072.
- Wolak, Jennifer, and Michael McDevitt. 2011. "The roots of the gender gap in political knowledge in adolescence." *Political Behavior* 33 (3): 505–533.
- Zaller, John. 1990. "Political awareness, elite opinion leadership, and the mass survey response." *Social Cognition* 8 (1): 125.
- Zaller, John. 1992. *The nature and origins of mass opinion*. Cambridge University Press.

Appendices

Online Appendix:

Let's Talk Politics – A Naive Approach for Measuring Political Sophistication

A Information on Discursive Sophistication Components	2
I Distribution of Word Counts in Open-Ended Responses	2
II Overview of Topic Proportions	3
III Discursive Sophistication Components	6
B Pre-Processing and Topic Model Specification	7
I PreText Analysis	7
II Robustness Checks for Varying Model Specifications	7
C Additional Information on Remaining Variables	10
I Item Selection and Recoding	10
II Variable Distributions – 2012 ANES	12
III Variable Distributions – 2016 ANES	14
IV Variable Distributions – 2015 YouGov	16
D Robustness Checks	17
I Controlling for Personality Characteristics	17
II Controlling for Individual Response Length	18
E Tables of Model Estimates	19
I Main Analyses	19
II Robustness Checks	26

Appendix A Detailed Information on Open-Ended Responses and Discursive Sophistication Components

I Distribution of Word Counts in Open-Ended Responses

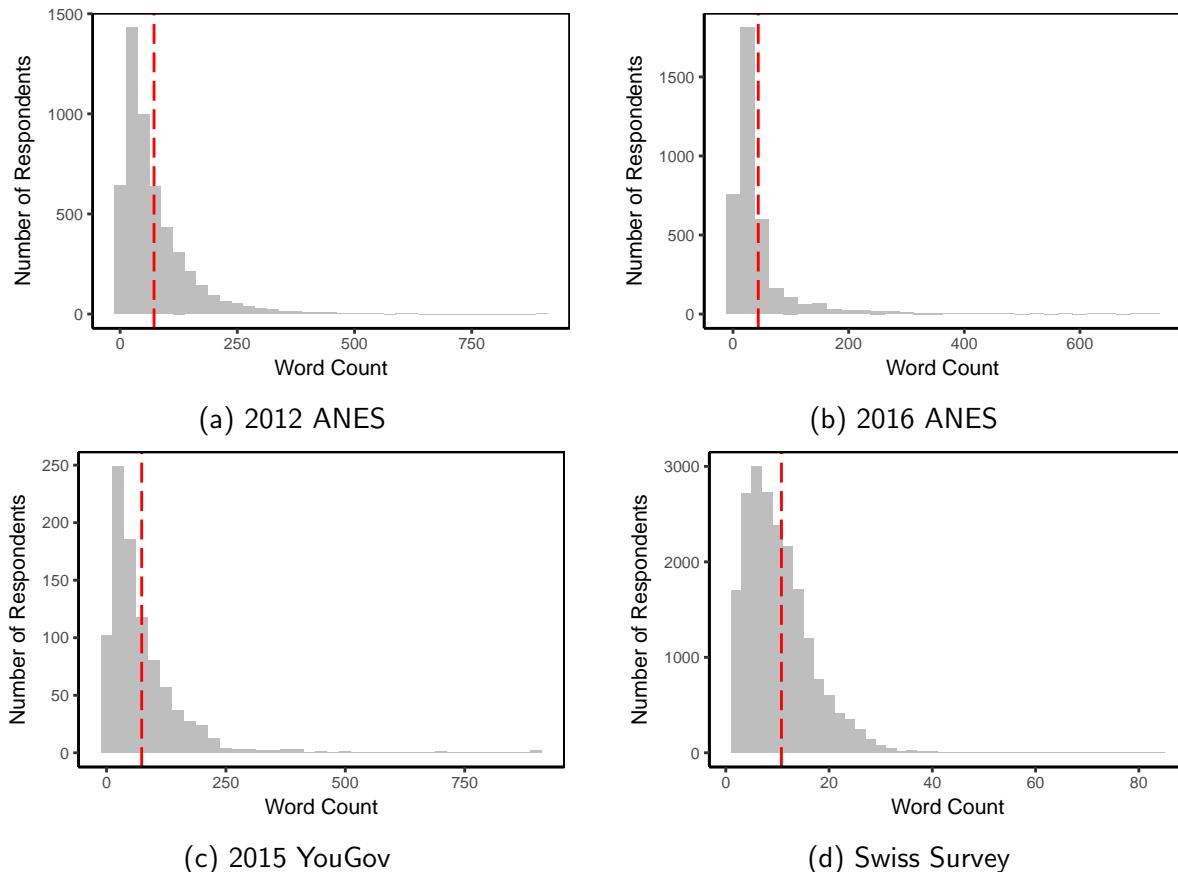


Figure A.1: Histograms of total word count in the collection of open-ended responses for each individual. The dashed red lines indicate the average response lengths in each survey.

II Overview of Topic Proportions

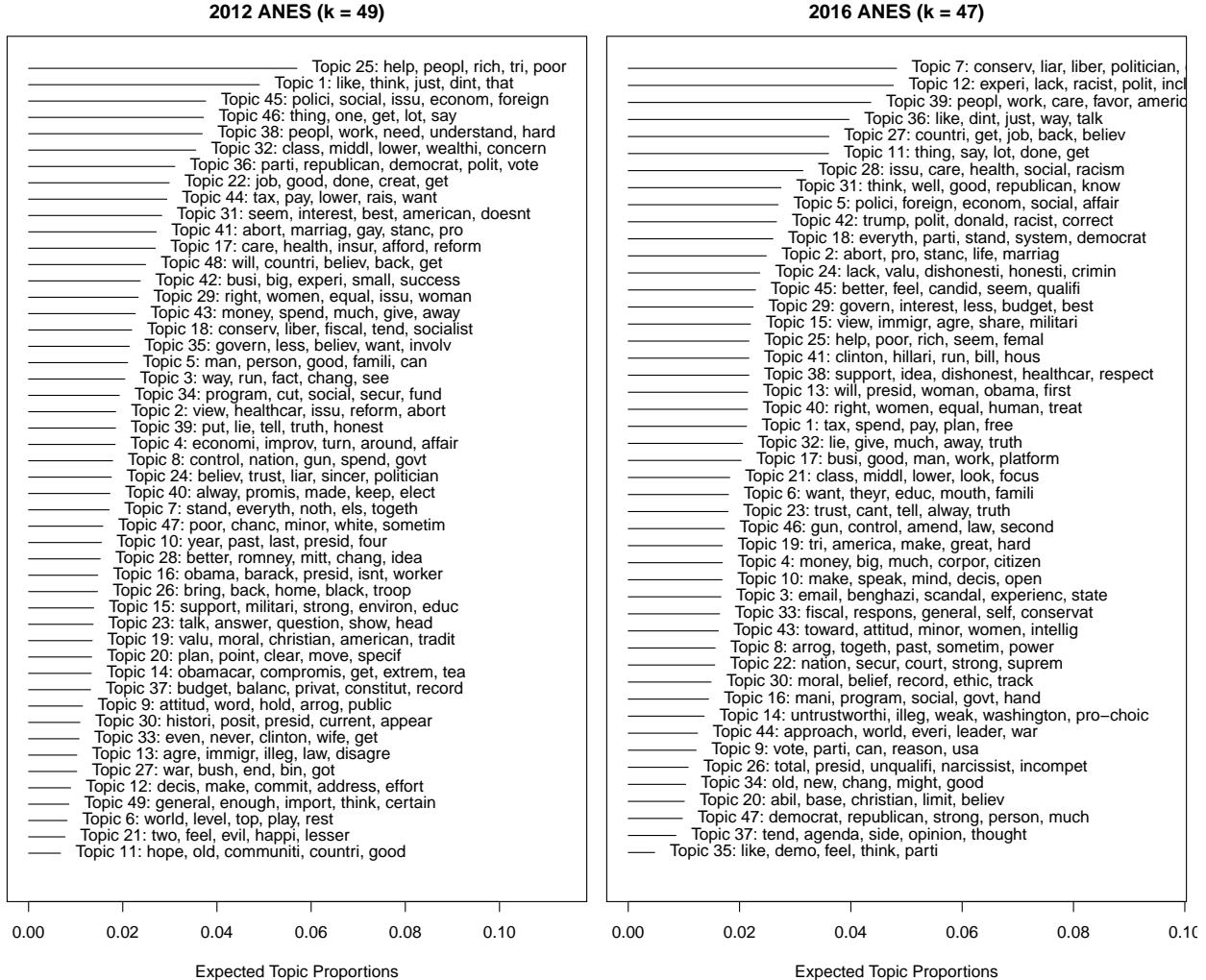


Figure A.2: Estimated topic proportions in the 2012 and 2016 ANES based on the structural topic model. See Appendix B for details on the model specification.



Figure A.3: Estimated topic proportions in the 2015 YouGov survey based on the structural topic model. See Appendix B for details on the model specification.

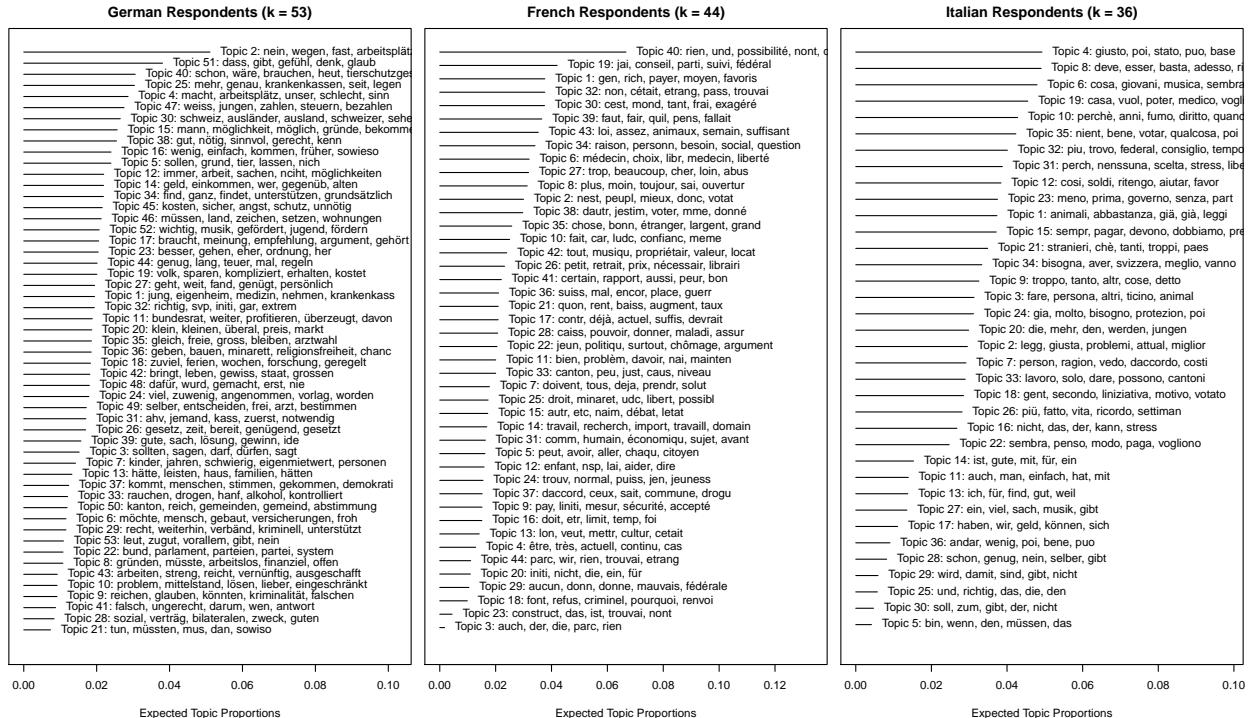


Figure A.4: Estimated topic proportions in the Swiss survey based on the structural topic model. See Appendix B for details on the model specification.

III Discursive Sophistication Components

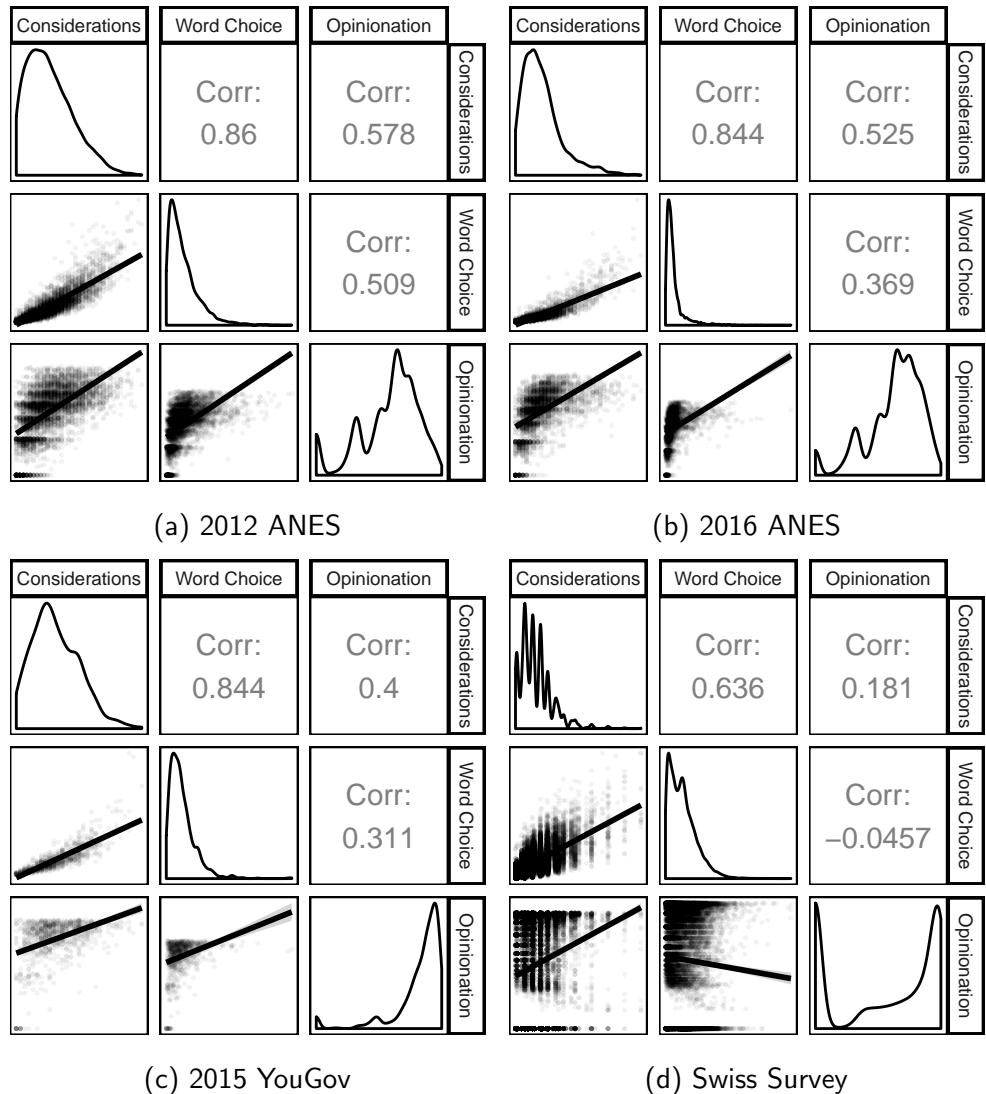


Figure A.5: Correlation matrix of individual components of discursive sophistication. The plots on the diagonal display univariate densities for each component. The panels in the lower triangular display the scatter plot of two measures as well as a linear fit.

Appendix B Pre-Processing and Topic Model Specification

I PreText Analysis

Two components of discursive sophistication (*considerations* and *word choice*) rely on quantities extracted from structural topic models (Roberts et al., 2014). As with any other text-as-data approach, a necessary first step before estimating the topic model is to pre-process the raw text and convert it into a document term matrix (DTM, see for example Manning et al., 2008). Common pre-processing procedures include stemming and lowercasing, as well as the removal of numbers, punctuation, stopwords, and infrequent terms. However, topic models and other unsupervised learning techniques can be sensitive to these pre-processing choices (c.f., Denny and Spirling, 2018). To address this issue, Denny and Spirling (2018) recommend that researchers compare DTMs under all possible pre-processing regimes. The authors propose *preText scores* as a measure to quantify the extent to which varying pre-processing regimes may yield unusual results compared to a baseline without any pre-processing.

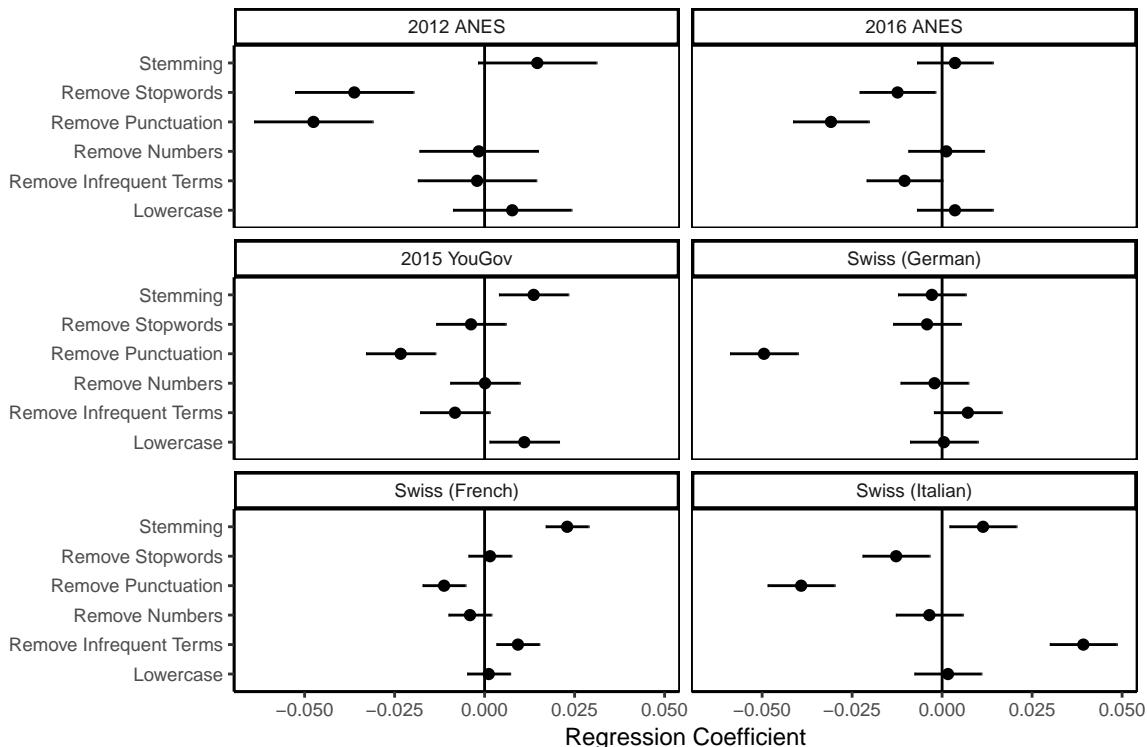


Figure B.1: PreText analysis of pre-processing decisions of open-ended responses across all datasets. Regression coefficients display the effects of each of the six pre-processing choices on the resulting preText score.

II Robustness Checks for Varying Model Specifications

Following the procedure outlined in Denny and Spirling (2018), Figure B.1 displays the results of a linear model regressing preText scores resulting from all possible pre-processing regimes on each

individual step for a random subset of 500 open-ended responses in each of the studies included in the analyses. Significant coefficients indicate that the topic model results may be sensitive to the respective pre-processing step. As such, removing stopwords and punctuation, as well as removing infrequent terms (at least in the 2016 ANES) might be problematic. Denny and Spirling (2018), however, emphasize that the most important consideration in choosing pre-processing steps are theoretical. Given that the purpose of the topic model is to extract considerations related to political preferences, there are strong theoretical reasons to remove stopwords and punctuation from open-ended responses as they do not contain any relevant content. Furthermore, I apply lowercasing and stemming of terms to reduce resulting document term matrix to a computationally more manageable size and since these pre-processing steps should not be influential according to the preText analysis.

It is less obvious from a theoretical perspective whether to remove infrequent terms from open-ended responses, although it is preferred in order to make the estimation of the discursive sophistication components computationally efficient. Since the preText analysis for the 2016 ANES suggests that this pre-processing step might be influential, I compare discursive sophistication for both alternative regimes below (c.f., Denny and Spirling, 2018). Before turning to this sensitivity check, however, I consider another crucial modeling choice when working with topic models: determining the total number of topics k to be estimated. For all analyses reported below, the number of topics was selected using the algorithm proposed by Lee and Mimno (2014) and implemented in the `stm` package in R (Roberts, Stewart, and Tingley, 2014).⁹

Figure B.2 examines whether the proposed measure of discursive sophistication is sensitive to the removal of infrequent terms as well as the chosen number of topics k . The y-axis depicts the preferred pre-processing regime including all steps discussed above while the x-axis plots results for alternative specifications. The panels on the left compare the preferred specification to discursive sophistication based on a reduced number of topics ($k = 20$). The middle panels additionally include infrequent terms instead of removing them.¹⁰ The panels on the right do not perform stemming as part of the pre-processing step. Across all panels, discursive sophistication scores are highly correlated and therefore insensitive to pre-processing choices and varying numbers of topics.

In summary, open-ended responses in the analyses reported in the main text are pre-processed by stemming and lowercasing, as well as the removing numbers, punctuation, stopwords, and infrequent terms (i.e., terms that appear in fewer than 10 responses).¹¹ While the results discussed in the manuscript are based on this preferred specification, the substantive results are robust for alternative pre-processing regimes or varying numbers of topics.

⁹I used measures for age, gender, education, party identification, as well as an interaction between education and party identification as covariates for topic prevalence. This variable selection—with the exception of including gender—is equivalent to the procedure model specification described in Roberts et al. (2014).

¹⁰Calculating discursive sophistication with large numbers of topics while including infrequent terms is computationally prohibitive.

¹¹Prior to applying these pre-processing steps, open-ended responses in the 2012 & 2016 ANES as well as the 2015 YouGov survey are cleaned by correcting spelling errors using an implementation of the Aspell spell-checking algorithm (www.aspell.net).

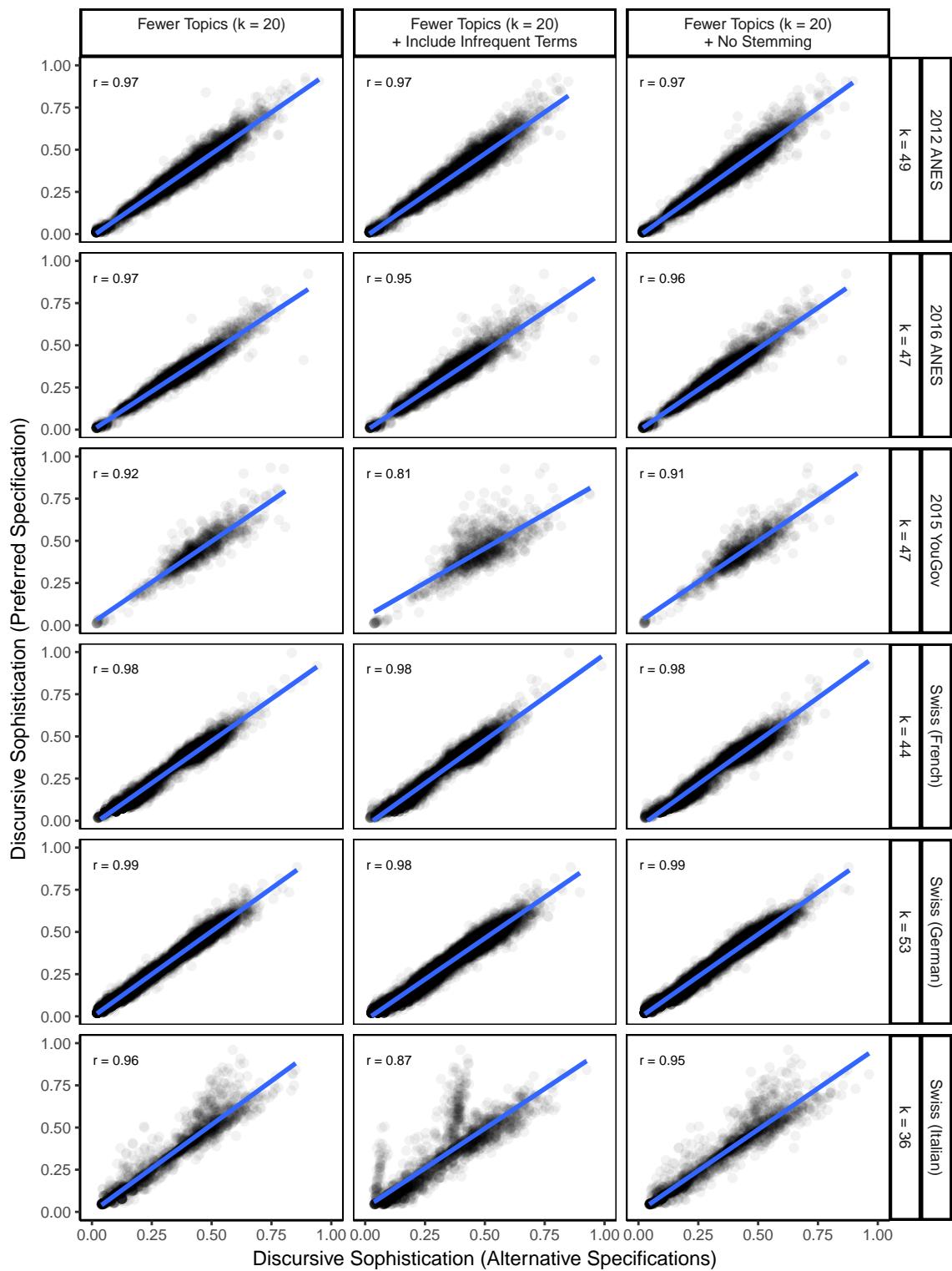


Figure B.2: Robustness of discursive sophistication measure for different pre-processing choices and topic model specifications.

Appendix C Additional Information on Remaining Variables

I Item Selection and Recoding

Conventional measures of political knowledge:

- *2012 ANES*: Additive index of correct responses to 5 knowledge items included in the pre-election wave (number of Presidential terms, size of budget deficit, length of Senate term, meaning of Medicare, federal government spending). ‘Don’t know’ responses are considered incorrect. Interviewer evaluations are based on the assessment of the respondent’s general level of information about politics recorded at the end of the pre-election wave.
- *2016 ANES*: Additive index of correct responses to 4 knowledge items included in the pre-election wave (length of Senate term, federal government spending, majority in House, majority in Senate). ‘Don’t know’ responses are considered incorrect. Interviewer evaluations are based on the assessment of the respondent’s general level of information about politics recorded at the end of the pre-election wave.
- *2015 YouGov Survey*: Additive index of correct responses to 8 knowledge items (Speaker of the House, meaning of TPP, Chair of Federal Reserve Board, current unemployment rate, Presidential veto override, meaning of Common Core, leading source of electricity in US, majority in Senate). ‘Don’t know’ responses are considered incorrect.

Dependent variables:

- *Turnout* (2012 & 2016 ANES): Dichotomous indicator, based on post-election wave.
- *Non-conventional participation* (2012 & 2016 ANES): Additive index of different forms of political engagement (participated in protest, signed petition, wore campaign button, wrote letter to public official).
- *Internal efficacy* (2012 & 2016 ANES): Sum of two agree/disagree items (politics too complicated, good understanding of political issues [reversed]).
- *External efficacy* (2012 & 2016 ANES): Sum of two agree/disagree items (public officials don’t care, people have no say about what the government does).
- *Information retrieval* (2015 YouGov Survey): Additive index of correct answers to 9 questions about the fictional disease described in the news article (symptoms: fatigue, headaches, diarrhea, joint pain, boils, warts, fever; virus spread; cure for the virus)
- *Candidate policy positions* (2012 & 2016 ANES): Placement of Republican and Democratic Presidential candidates on 7-point scale (ideology, government spending, defense spending, insurance policy, job guarantee, aid to Blacks, environment vs jobs).

Control variables:

- *Gender* (2012 & 2016 ANES, 2015 YouGov Survey): Dichotomous indicator for female respondents.
- *Wordsum vocabulary scores* (2012 & 2016 ANES): Modified version of the GSS wordsum vocabulary test consisting of 10 terms.
- *Media exposure* (2012 & 2016 ANES): Additive index of the frequency of weekly exposure to various political information sources such as newspapers or TV news (2012 ANES). In the 2016 ANES, it only consists of a single item measuring the number of days in the past week the respondent has spent watching/reading/listening news on any media.
- *Political discussion frequency* (2012 & 2016 ANES): Self-reported count of days in the past week spent discussing politics with family or friends.
- *College education* (2012 & 2016 ANES, 2015 YouGov Survey): Dichotomous indicator for Bachelor's degree or higher.
- *Family/Household income* (2012 & 2016 ANES, 2015 YouGov Survey): Self-reported household income categories.
- *Age* (2012 & 2016 ANES, 2015 YouGov Survey): Logged age in years.
- *Race* (2012 & 2016 ANES, 2015 YouGov Survey): Dichotomous indicator for black non-Hispanic vs. others.
- *Church attendance* (2012 & 2016 ANES, 2015 YouGov Survey): Six-category indicator of the frequency of church attendance.
- *Survey Mode* (2012 & 2016 ANES): Dichotomous indicator for face-to-face vs. online samples of the ANES surveys.
- *Personality characteristics* (2012 & 2016 ANES): Measures of extraversion and being reserved, part of the Ten Item Personality Inventory (TIPI) measuring the “Big Five” personality traits.
- *Response length* (2012 & 2016 ANES): Logged number of words in the collection of open-ended responses by each individual.

II Variable Distributions – 2012 ANES

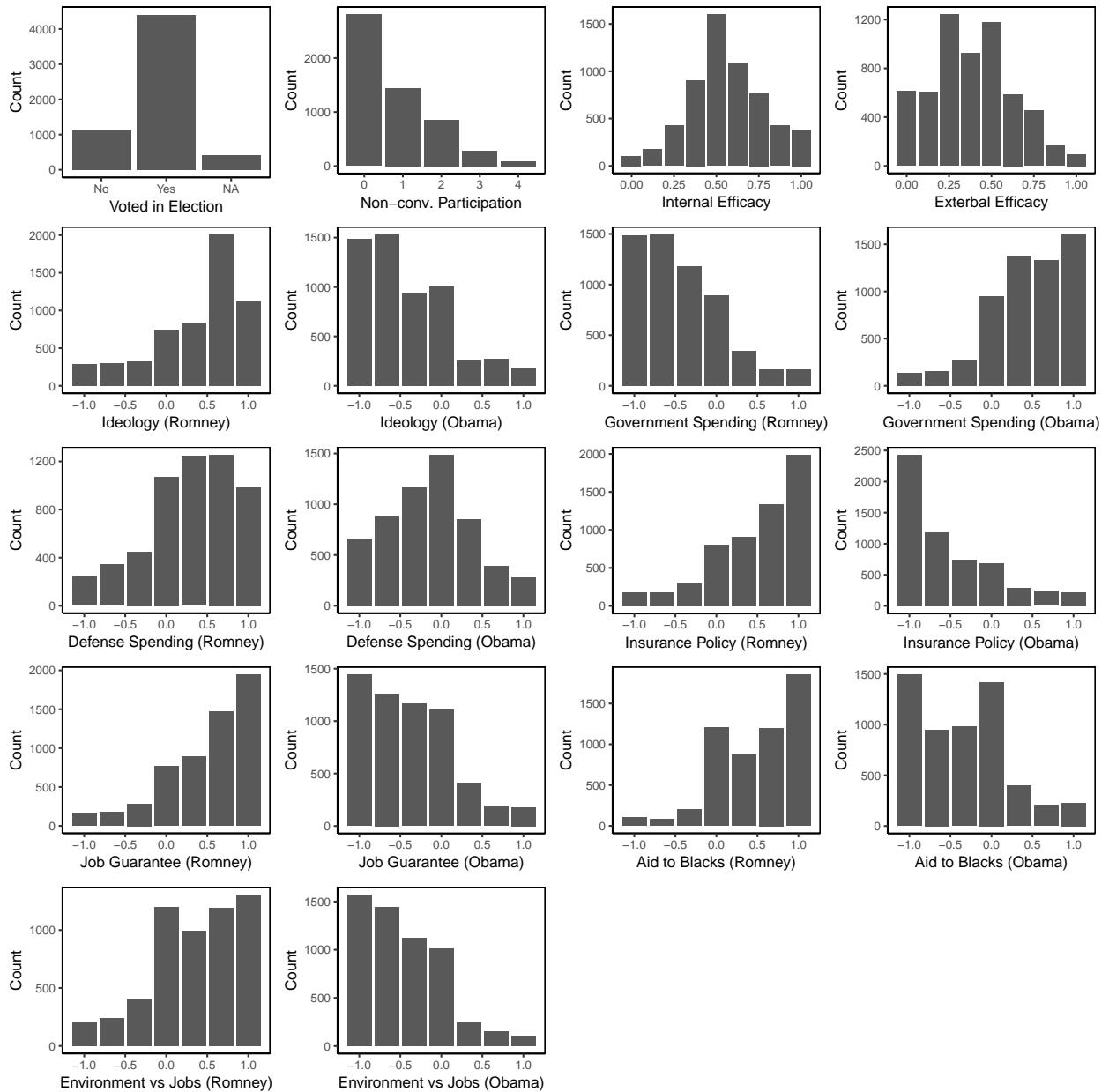


Figure C.1: Histograms of dependent variables included in 2012 ANES.

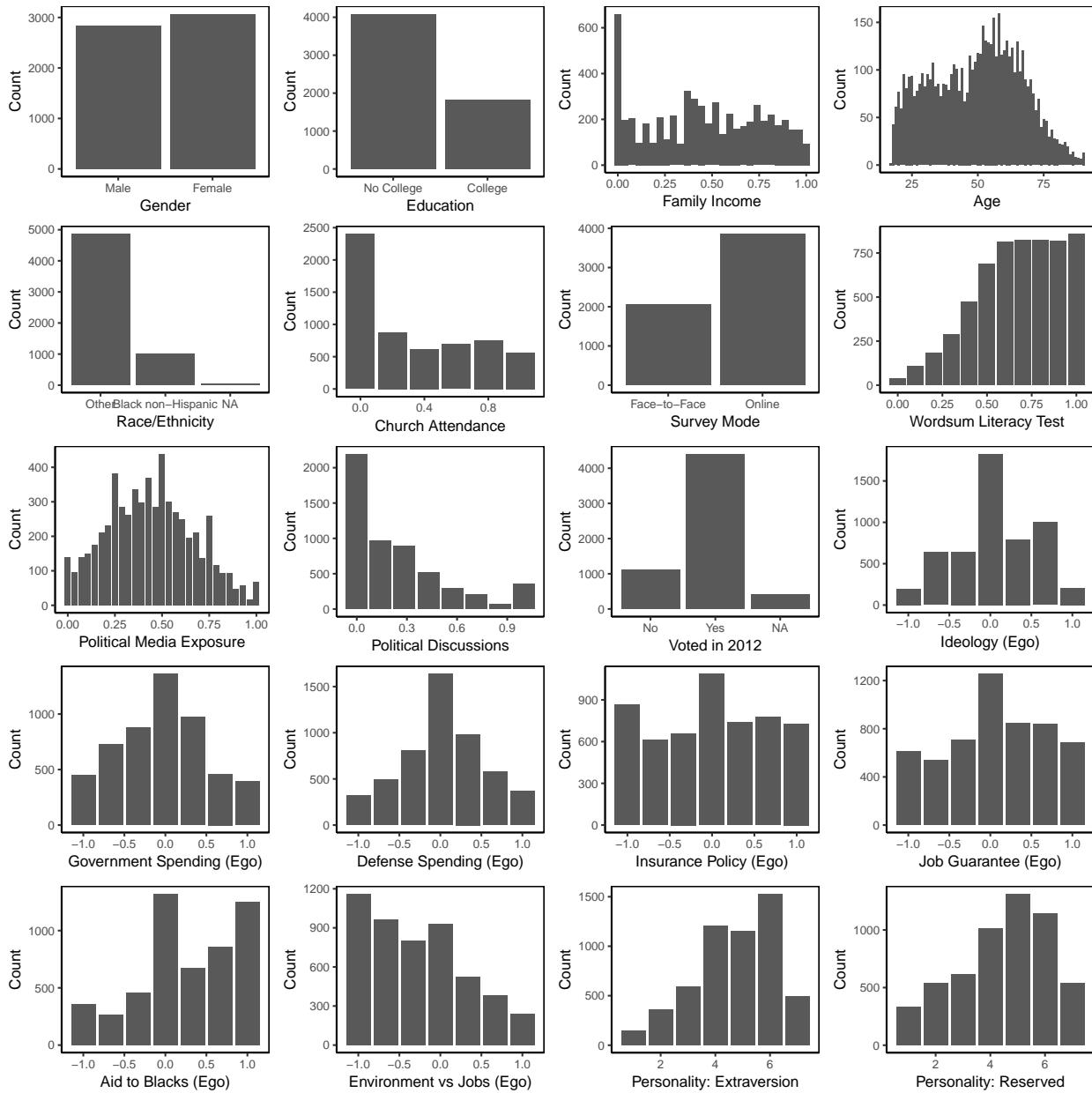


Figure C.2: Histograms of independent variables included in 2012 ANES.

III Variable Distributions – 2016 ANES

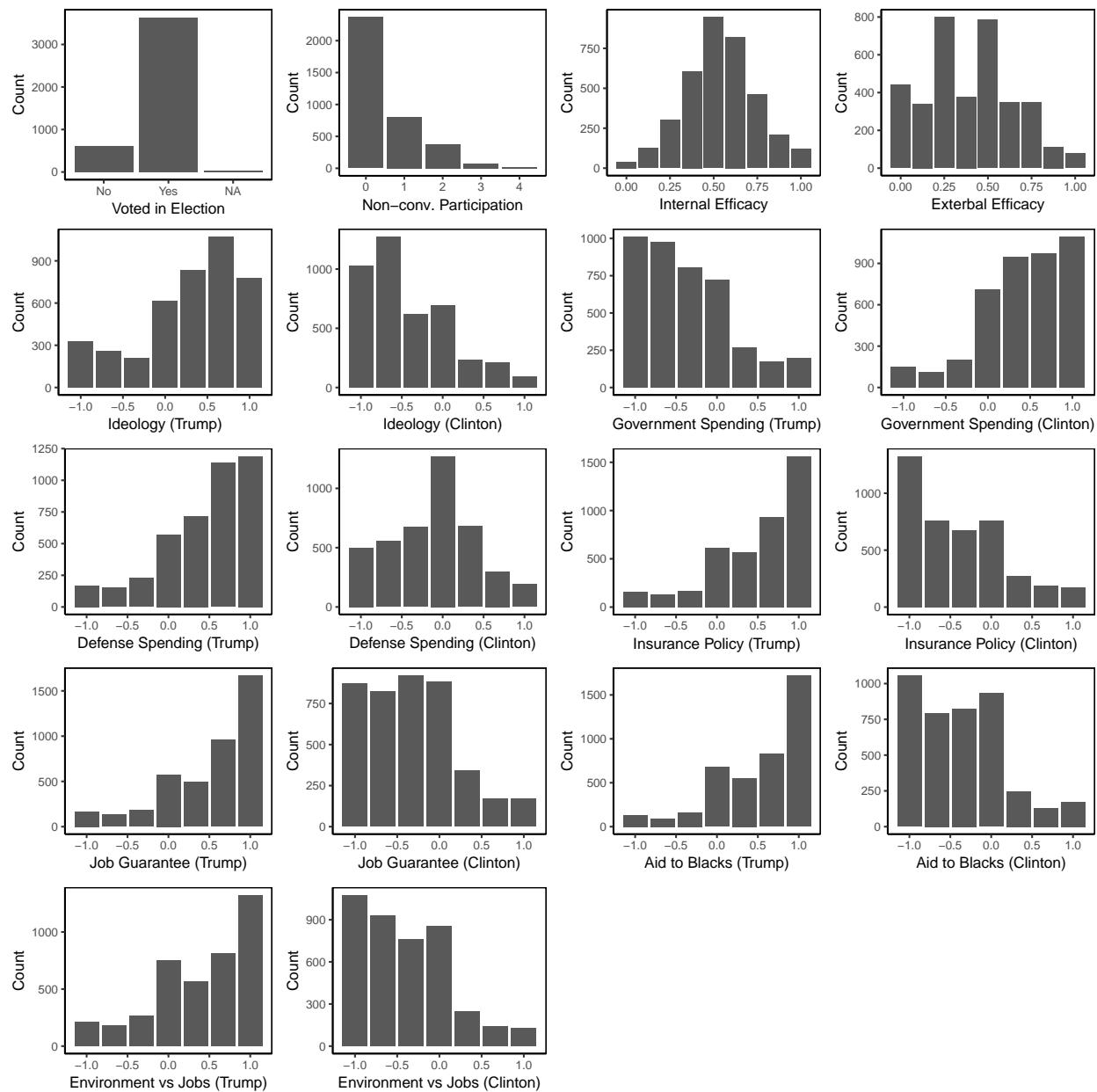


Figure C.3: Histograms of dependent variables included in 2016 ANES.

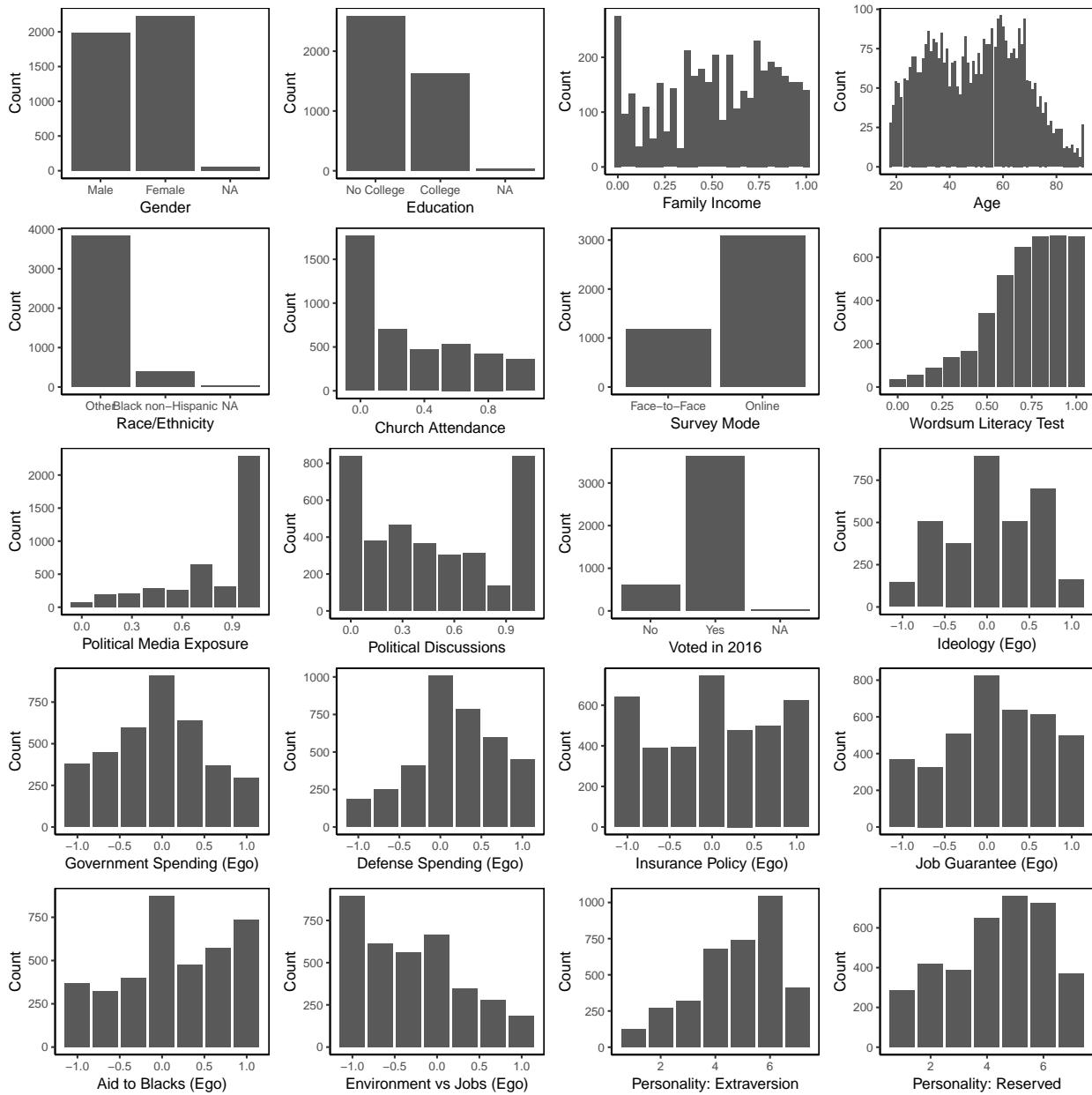


Figure C.4: Histograms of independent variables included in 2016 ANES.

IV Variable Distributions – 2015 YouGov

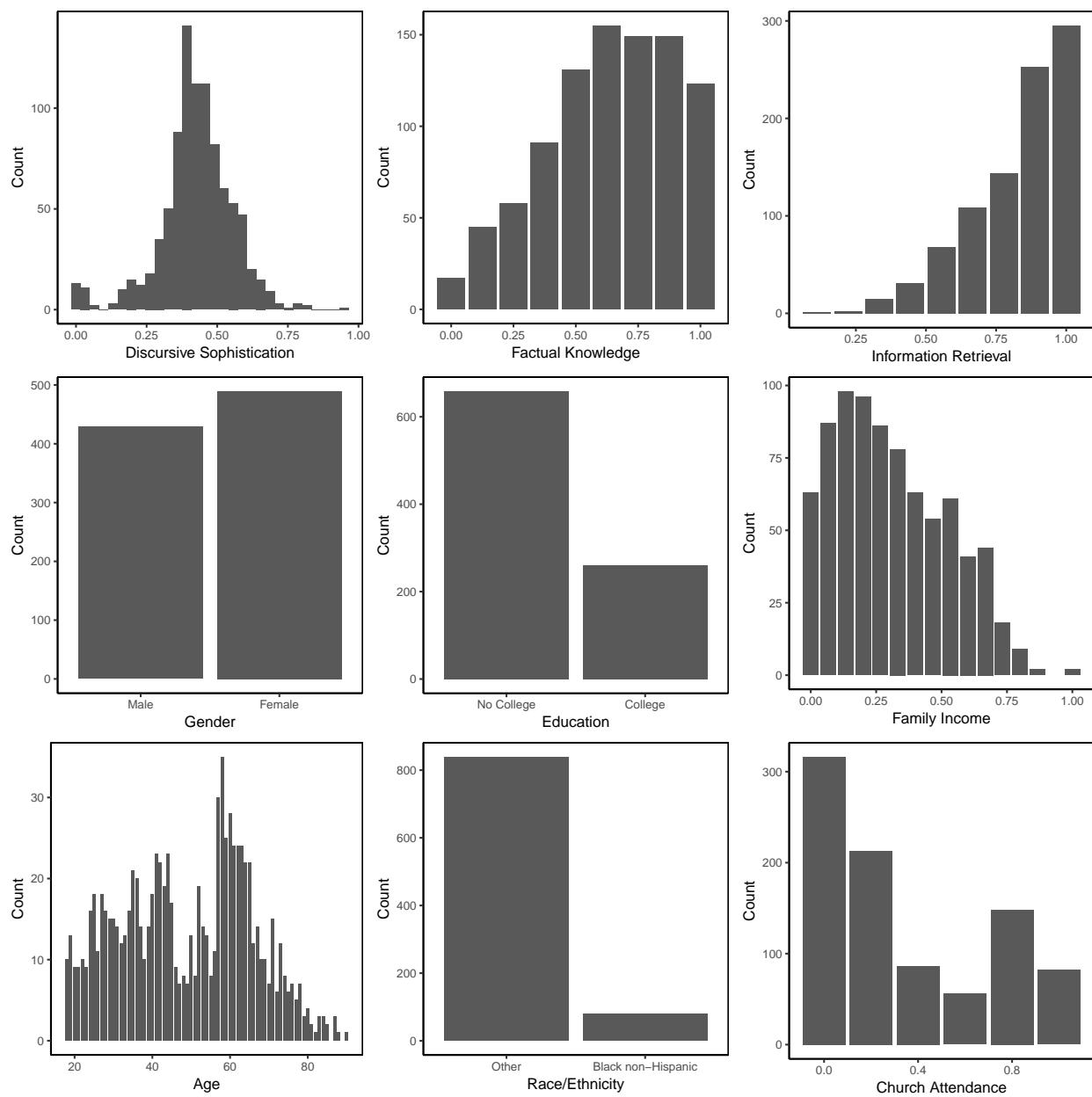


Figure C.5: Histogram of variables included in 2015 YouGov survey.

Appendix D Robustness Checks

I Controlling for Personality Characteristics

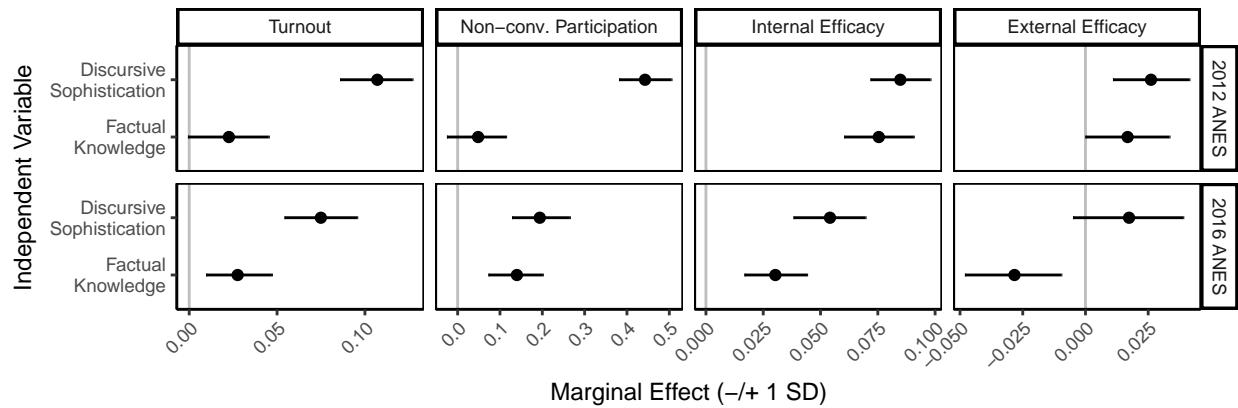


Figure D.1: Effects of sophistication on internal efficacy, external efficacy, non-conventional participation, and turnout in the 2012 and 2016 ANES. For each dependent variable, the figure displays the change in expected values after increasing each sophistication measure from -1 to +1 standard deviation from its mean (including 95% confidence intervals). Model estimates are based on logistic regression (turnout) or OLS (internal efficacy, external efficacy, non-conventional participation). Both sophistication measure are included simultaneously while controlling for gender, education, income, age, race, church attendance, survey mode, Wordsum vocabulary scores, as well as personality characteristics (extraversion and being reserved). Full model results are presented in the appendix, Tables 9 and 10.

II Controlling for Individual Response Length

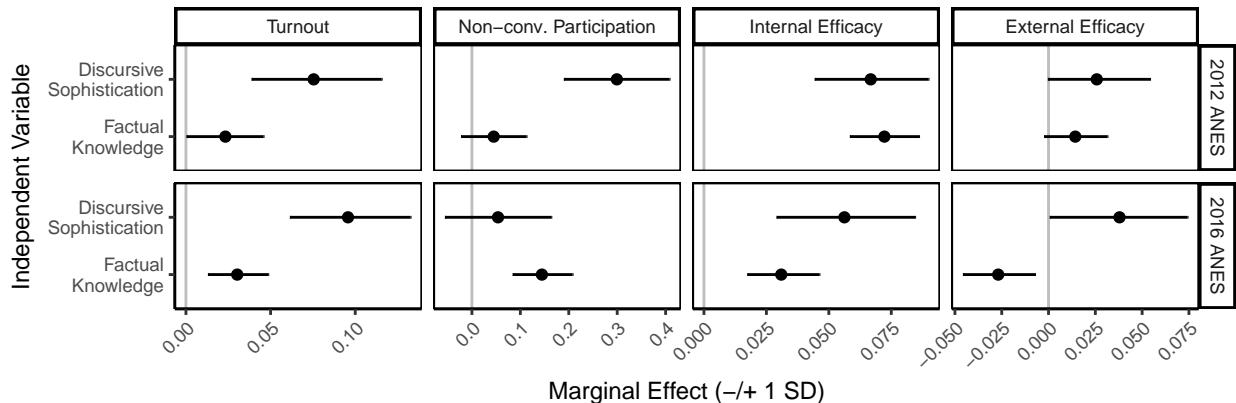


Figure D.2: Effects of sophistication on internal efficacy, external efficacy, non-conventional participation, and turnout in the 2012 and 2016 ANES. For each dependent variable, the figure displays the change in expected values after increasing each sophistication measure from -1 to +1 standard deviation from its mean (including 95% confidence intervals). Model estimates are based on logistic regression (turnout) or OLS (internal efficacy, external efficacy, non-conventional participation). Both sophistication measure are included simultaneously while controlling for gender, education, income, age, race, church attendance, survey mode, Wordsum vocabulary scores, as well as the logged word count across open-ended responses. Full model results are presented in the appendix, Tables 11 and 12.

Appendix E Tables of Model Estimates

I Main Analyses

Table 2: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy in the 2012 ANES. Standard errors in parentheses. Estimates are used for Figure 2 in the main text.

	<i>Dependent variable:</i>			
	Turnout	Participation	Internal Efficacy	External Efficacy
Discursive Soph.	2.921 (0.299)	1.440 (0.098)	0.278 (0.020)	0.084 (0.024)
Factual Knowledge	0.432 (0.218)	0.099 (0.075)	0.158 (0.016)	0.032 (0.018)
Female	0.086 (0.085)	-0.067 (0.028)	-0.053 (0.006)	0.016 (0.007)
College Degree	0.350 (0.112)	0.159 (0.034)	0.022 (0.007)	0.035 (0.008)
Family Income	0.947 (0.156)	0.022 (0.052)	0.010 (0.011)	0.016 (0.013)
Age (log)	0.988 (0.105)	0.102 (0.038)	-0.006 (0.008)	-0.014 (0.009)
African American	0.910 (0.123)	0.096 (0.038)	0.066 (0.008)	0.082 (0.009)
Church Attendance	0.752 (0.129)	0.112 (0.040)	0.010 (0.008)	0.048 (0.010)
Mode: Online	0.530 (0.094)	0.227 (0.033)	0.017 (0.007)	-0.039 (0.008)
Wordsum Score	0.638 (0.219)	0.403 (0.076)	0.092 (0.016)	0.013 (0.019)
Constant	-5.019 (0.401)	-0.598 (0.145)	0.326 (0.030)	0.352 (0.035)
Observations	4,716	4,692	4,996	4,985
R ²		0.124	0.161	0.043
Log Likelihood	-1,868.199			

Table 3: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy in the 2016 ANES. Standard errors in parentheses. Estimates are used for Figure 2 in the main text.

	<i>Dependent variable:</i>			
	Turnout	Participation	Internal Efficacy	External Efficacy
Discursive Soph.	3.891 (0.487)	0.792 (0.131)	0.229 (0.031)	0.076 (0.041)
Factual Knowledge	0.730 (0.219)	0.269 (0.058)	0.058 (0.014)	-0.051 (0.018)
Female	0.177 (0.114)	0.061 (0.029)	-0.059 (0.007)	-0.003 (0.009)
College Degree	0.554 (0.141)	0.092 (0.033)	0.058 (0.008)	0.057 (0.010)
Family Income	0.487 (0.207)	-0.077 (0.055)	0.020 (0.013)	0.062 (0.017)
Age (log)	0.840 (0.139)	-0.110 (0.038)	0.020 (0.009)	-0.005 (0.012)
African American	1.126 (0.237)	0.097 (0.051)	0.057 (0.012)	-0.020 (0.016)
Church Attendance	1.067 (0.191)	-0.175 (0.043)	-0.006 (0.010)	0.080 (0.014)
Mode: Online	0.192 (0.136)	0.110 (0.036)	0.069 (0.009)	-0.016 (0.011)
Wordsum Score	0.931 (0.275)	0.402 (0.079)	0.103 (0.019)	0.017 (0.025)
Constant	-4.276 (0.535)	0.213 (0.151)	0.231 (0.036)	0.338 (0.048)
Observations	3,486	3,039	3,037	3,038
R ²		0.062	0.147	0.043
Log Likelihood	-1,085.646			

Table 4: Effects of sophistication on information retrieval in the 2015 YouGov study. Standard errors in parentheses. Estimates are used for Figure 3 in the main text.

	<i>Dependent variable:</i> Information Retrieval
Discursive Soph.	0.247 (0.045)
Factual Knowledge	0.133 (0.027)
Female	0.044 (0.012)
College Degree	0.015 (0.014)
Family Income	−0.062 (0.030)
Age (log)	0.032 (0.016)
African American	−0.026 (0.021)
Church Attendance	−0.059 (0.017)
Constant	0.534 (0.061)
Observations	792
R ²	0.136

Table 5: Error variance reduction in candidate placements on multiple issues in the 2012 ANES. Estimates are used for Figure 4 in the main text.

Policy Position	Candidate	Independent Var.	$E[\gamma]$	$sd(\gamma)$	95% Cred. Int.	\hat{R}
Ideology	Romney	Discursive Soph.	-0.71	0.07	(-0.856; -0.562)	1.00
Ideology	Romney	Factual Knowl.	-0.65	0.06	(-0.755; -0.538)	1.00
Ideology	Obama	Discursive Soph.	-0.41	0.07	(-0.544; -0.264)	1.00
Ideology	Obama	Factual Knowl.	-0.58	0.06	(-0.69; -0.47)	1.00
Government Spending	Romney	Discursive Soph.	-0.34	0.07	(-0.482; -0.205)	1.00
Government Spending	Romney	Factual Knowl.	-0.41	0.06	(-0.52; -0.303)	1.00
Government Spending	Obama	Discursive Soph.	-0.54	0.07	(-0.68; -0.398)	1.00
Government Spending	Obama	Factual Knowl.	-0.47	0.06	(-0.582; -0.354)	1.00
Defense Spending	Romney	Discursive Soph.	-0.18	0.08	(-0.326; -0.031)	1.00
Defense Spending	Romney	Factual Knowl.	-0.33	0.06	(-0.443; -0.223)	1.00
Defense Spending	Obama	Discursive Soph.	-0.17	0.08	(-0.318; -0.028)	1.00
Defense Spending	Obama	Factual Knowl.	-0.19	0.05	(-0.287; -0.086)	1.00
Insurance Policy	Romney	Discursive Soph.	-0.26	0.07	(-0.398; -0.12)	1.00
Insurance Policy	Romney	Factual Knowl.	-0.22	0.05	(-0.326; -0.116)	1.00
Insurance Policy	Obama	Discursive Soph.	-0.33	0.07	(-0.476; -0.194)	1.00
Insurance Policy	Obama	Factual Knowl.	-0.41	0.06	(-0.515; -0.297)	1.00
Job Guarantee	Romney	Discursive Soph.	-0.59	0.08	(-0.734; -0.436)	1.00
Job Guarantee	Romney	Factual Knowl.	-0.33	0.06	(-0.439; -0.223)	1.00
Job Guarantee	Obama	Discursive Soph.	-0.35	0.07	(-0.498; -0.215)	1.00
Job Guarantee	Obama	Factual Knowl.	-0.20	0.05	(-0.304; -0.091)	1.00
Aid to Blacks	Romney	Discursive Soph.	-0.22	0.07	(-0.368; -0.069)	1.00
Aid to Blacks	Romney	Factual Knowl.	-0.27	0.05	(-0.369; -0.163)	1.00
Aid to Blacks	Obama	Discursive Soph.	-0.21	0.07	(-0.36; -0.07)	1.00
Aid to Blacks	Obama	Factual Knowl.	-0.20	0.06	(-0.312; -0.095)	1.00
Environment vs Jobs	Romney	Discursive Soph.	-0.29	0.08	(-0.438; -0.138)	1.00
Environment vs Jobs	Romney	Factual Knowl.	-0.34	0.06	(-0.456; -0.227)	1.00
Environment vs Jobs	Obama	Discursive Soph.	-0.34	0.08	(-0.484; -0.181)	1.00
Environment vs Jobs	Obama	Factual Knowl.	-0.35	0.06	(-0.465; -0.233)	1.00

Table 6: Error variance reduction in candidate placements on multiple issues in the 2016 ANES. Estimates are used for Figure 4 in the main text.

Policy Position	Candidate	Independent Var.	$E[\gamma]$	$sd(\gamma)$	95% Cred. Int.	\hat{R}
Ideology	Trump	Discursive Soph.	-0.77	0.11	(-0.984; -0.55)	1.00
Ideology	Trump	Factual Knowl.	-0.16	0.06	(-0.272; -0.057)	1.00
Ideology	Clinton	Discursive Soph.	-0.52	0.10	(-0.71; -0.313)	1.00
Ideology	Clinton	Factual Knowl.	-0.15	0.05	(-0.255; -0.048)	1.00
Government Spending	Trump	Discursive Soph.	-0.62	0.10	(-0.835; -0.424)	1.00
Government Spending	Trump	Factual Knowl.	-0.09	0.05	(-0.19; 0.013)	1.00
Government Spending	Clinton	Discursive Soph.	-1.00	0.11	(-1.213; -0.785)	1.00
Government Spending	Clinton	Factual Knowl.	-0.18	0.05	(-0.284; -0.079)	1.00
Defense Spending	Trump	Discursive Soph.	-0.37	0.10	(-0.563; -0.171)	1.00
Defense Spending	Trump	Factual Knowl.	-0.08	0.05	(-0.183; 0.021)	1.00
Defense Spending	Clinton	Discursive Soph.	-0.54	0.11	(-0.744; -0.319)	1.00
Defense Spending	Clinton	Factual Knowl.	-0.04	0.05	(-0.141; 0.061)	1.00
Insurance Policy	Trump	Discursive Soph.	-0.79	0.10	(-0.991; -0.579)	1.00
Insurance Policy	Trump	Factual Knowl.	-0.06	0.05	(-0.168; 0.04)	1.00
Insurance Policy	Clinton	Discursive Soph.	-0.74	0.11	(-0.959; -0.521)	1.00
Insurance Policy	Clinton	Factual Knowl.	-0.14	0.05	(-0.241; -0.033)	1.00
Job Guarantee	Trump	Discursive Soph.	-0.86	0.10	(-1.061; -0.671)	1.00
Job Guarantee	Trump	Factual Knowl.	-0.06	0.05	(-0.157; 0.043)	1.00
Job Guarantee	Clinton	Discursive Soph.	-0.76	0.10	(-0.965; -0.546)	1.00
Job Guarantee	Clinton	Factual Knowl.	-0.04	0.05	(-0.134; 0.059)	1.00
Aid to Blacks	Trump	Discursive Soph.	-0.66	0.10	(-0.862; -0.466)	1.00
Aid to Blacks	Trump	Factual Knowl.	-0.00	0.05	(-0.099; 0.097)	1.00
Aid to Blacks	Clinton	Discursive Soph.	-0.47	0.10	(-0.663; -0.277)	1.00
Aid to Blacks	Clinton	Factual Knowl.	-0.06	0.05	(-0.158; 0.045)	1.00
Environment vs Jobs	Trump	Discursive Soph.	-0.65	0.11	(-0.861; -0.443)	1.00
Environment vs Jobs	Trump	Factual Knowl.	-0.18	0.05	(-0.281; -0.072)	1.00
Environment vs Jobs	Clinton	Discursive Soph.	-0.76	0.10	(-0.956; -0.554)	1.00
Environment vs Jobs	Clinton	Factual Knowl.	-0.24	0.05	(-0.348; -0.142)	1.00

Table 7: Effects of gender on discursive sophistication and factual knowledge in the 2012 ANES and 2016 ANES. Standard errors in parentheses. Estimates are used for Figure 7 in the main text.

	<i>Dependent variable:</i>			
	Discursive Sophistication		Factual Knowledge	
	2012 ANES	2016 ANES	2012 ANES	2016 ANES
Female	0.001 (0.004)	-0.002 (0.004)	-0.050 (0.006)	-0.029 (0.009)
Media Exposure	0.042 (0.010)	0.034 (0.008)	0.067 (0.014)	0.056 (0.019)
Discussion Frequency	0.121 (0.008)	0.058 (0.006)	0.062 (0.010)	0.078 (0.013)
College Degree	0.070 (0.005)	0.045 (0.004)	0.091 (0.007)	0.045 (0.010)
Family Income	0.054 (0.008)	0.043 (0.007)	0.116 (0.010)	0.089 (0.017)
Age (log)	0.017 (0.006)	0.003 (0.005)	0.081 (0.008)	0.092 (0.013)
African American	-0.010 (0.006)	-0.026 (0.007)	-0.085 (0.008)	-0.004 (0.016)
Church Attendance	0.004 (0.006)	0.005 (0.006)	-0.003 (0.008)	-0.006 (0.013)
Mode: Online	-0.044 (0.005)	-0.110 (0.004)	0.092 (0.006)	0.085 (0.010)
Constant	0.199 (0.022)	0.284 (0.020)	0.195 (0.030)	0.009 (0.046)
Observations	4,690	3,119	4,690	3,108
R ²	0.178	0.265	0.282	0.110

Table 8: Effects of gender on discursive sophistication and factual knowledge in the 2015 YouGov Study. Standard errors in parentheses. Estimates are used for Figure 7 in the main text.

	<i>Dependent variable:</i>	
	Discursive Sophistication	Factual Knowledge
Female	0.008 (0.009)	-0.157 (0.016)
College Degree	0.027 (0.011)	0.108 (0.019)
Family Income	0.003 (0.024)	0.233 (0.041)
Age (log)	0.040 (0.013)	0.135 (0.021)
African American	-0.064 (0.017)	-0.086 (0.029)
Church Attendance	-0.008 (0.013)	-0.051 (0.023)
Constant	0.272 (0.049)	0.125 (0.083)
Observations	792	792
R ²	0.043	0.272

II Robustness Checks

Table 9: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy controlling for personality characteristics in the 2012 ANES. Standard errors in parentheses. Estimates are used for Figure D.1 in the appendix.

	<i>Dependent variable:</i>			
	Turnout	Participation	Internal Efficacy	External Efficacy
Discursive Soph.	2.912 (0.300)	1.419 (0.098)	0.272 (0.021)	0.085 (0.025)
Factual Knowledge	0.434 (0.218)	0.105 (0.075)	0.165 (0.016)	0.036 (0.019)
Female	0.088 (0.085)	-0.078 (0.028)	-0.059 (0.006)	0.009 (0.007)
College Degree	0.352 (0.112)	0.160 (0.034)	0.020 (0.007)	0.033 (0.008)
Family Income	0.941 (0.157)	-0.002 (0.052)	0.006 (0.011)	0.010 (0.013)
Age (log)	0.981 (0.106)	0.102 (0.039)	-0.003 (0.008)	-0.017 (0.010)
African American	0.902 (0.124)	0.101 (0.038)	0.067 (0.008)	0.078 (0.010)
Church Attendance	0.732 (0.130)	0.104 (0.040)	0.006 (0.008)	0.044 (0.010)
Mode: Online	0.541 (0.096)	0.249 (0.033)	0.022 (0.007)	-0.033 (0.008)
Wordsum Score	0.649 (0.220)	0.409 (0.076)	0.092 (0.016)	0.021 (0.019)
Extraversion	0.029 (0.030)	0.032 (0.010)	0.009 (0.002)	0.020 (0.002)
Reserved	0.009 (0.026)	-0.023 (0.009)	-0.005 (0.002)	0.002 (0.002)
Constant	-5.181 (0.449)	-0.645 (0.158)	0.296 (0.034)	0.255 (0.039)
Observations	4,691	4,685	4,691	4,679
R ²		0.129	0.171	0.058
Log Likelihood	-1,861.790			

Table 10: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy controlling for personality characteristics in the 2016 ANES. Standard errors in parentheses. Estimates are used for Figure D.1 in the appendix.

	<i>Dependent variable:</i>			
	Turnout	Participation	Internal Efficacy	External Efficacy
Discursive Soph.	3.936 (0.543)	0.760 (0.132)	0.213 (0.031)	0.070 (0.042)
Factual Knowledge	0.717 (0.241)	0.260 (0.059)	0.056 (0.014)	-0.052 (0.019)
Female	0.172 (0.125)	0.046 (0.029)	-0.064 (0.007)	-0.004 (0.009)
College Degree	0.616 (0.155)	0.085 (0.033)	0.056 (0.008)	0.055 (0.010)
Family Income	0.471 (0.230)	-0.085 (0.055)	0.017 (0.013)	0.060 (0.017)
Age (log)	0.894 (0.153)	-0.119 (0.039)	0.018 (0.009)	-0.010 (0.012)
African American	1.228 (0.267)	0.102 (0.052)	0.057 (0.012)	-0.026 (0.016)
Church Attendance	0.930 (0.205)	-0.172 (0.043)	-0.007 (0.010)	0.078 (0.014)
Mode: Online	0.199 (0.148)	0.111 (0.036)	0.069 (0.009)	-0.015 (0.011)
Wordsum Score	0.999 (0.311)	0.436 (0.080)	0.114 (0.019)	0.022 (0.025)
Extraversion	0.090 (0.041)	0.023 (0.010)	0.011 (0.002)	0.012 (0.003)
Reserved	0.005 (0.037)	-0.013 (0.009)	-0.002 (0.002)	0.005 (0.003)
Constant	-4.948 (0.646)	0.196 (0.163)	0.199 (0.039)	0.277 (0.051)
Observations	2,999	3,011	3,012	3,012
R ²		0.065	0.155	0.047
Log Likelihood	-907.430			

Table 11: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy controlling for individual response length in the 2012 ANES. Standard errors in parentheses. Estimates are used for Figure D.2 in the appendix.

	<i>Dependent variable:</i>			
	Turnout	Participation	Internal Efficacy	External Efficacy
Discursive Soph.	2.053 (0.543)	0.967 (0.186)	0.215 (0.039)	0.082 (0.045)
Factual Knowledge	0.426 (0.218)	0.096 (0.075)	0.158 (0.016)	0.032 (0.018)
Female	0.067 (0.085)	-0.076 (0.028)	-0.054 (0.006)	0.016 (0.007)
College Degree	0.370 (0.113)	0.167 (0.034)	0.023 (0.007)	0.035 (0.008)
Family Income	0.948 (0.156)	0.026 (0.052)	0.011 (0.011)	0.016 (0.013)
Age (log)	0.976 (0.106)	0.094 (0.039)	-0.007 (0.008)	-0.014 (0.009)
African American	0.896 (0.123)	0.088 (0.038)	0.064 (0.008)	0.082 (0.009)
Church Attendance	0.758 (0.129)	0.113 (0.040)	0.010 (0.008)	0.048 (0.010)
Mode: Online	0.574 (0.097)	0.246 (0.034)	0.020 (0.007)	-0.039 (0.008)
Wordsum Score	0.626 (0.219)	0.396 (0.076)	0.091 (0.016)	0.013 (0.019)
Word Count (log)	0.970 (0.507)	0.539 (0.180)	0.071 (0.037)	0.003 (0.044)
Constant	-5.237 (0.418)	-0.710 (0.150)	0.311 (0.031)	0.352 (0.036)
Observations	4,716	4,692	4,996	4,985
R ²		0.126	0.162	0.043
Log Likelihood	-1,866.366			

Table 12: Effects of sophistication on turnout, non-conventional participation, internal efficacy, and external efficacy controlling for individual response length in the 2016 ANES. Standard errors in parentheses. Estimates are used for Figure D.2 in the appendix.

	<i>Dependent variable:</i>			
	Turnout	Participation	Internal Efficacy	External Efficacy
Discursive Soph.	4.762 (0.843)	0.224 (0.231)	0.222 (0.055)	0.153 (0.073)
Factual Knowledge	0.727 (0.219)	0.269 (0.058)	0.058 (0.014)	-0.051 (0.018)
Female	0.189 (0.114)	0.054 (0.029)	-0.059 (0.007)	-0.002 (0.009)
College Degree	0.547 (0.141)	0.096 (0.033)	0.058 (0.008)	0.056 (0.010)
Family Income	0.487 (0.207)	-0.074 (0.055)	0.020 (0.013)	0.062 (0.017)
Age (log)	0.842 (0.139)	-0.111 (0.038)	0.020 (0.009)	-0.005 (0.012)
African American	1.137 (0.237)	0.090 (0.051)	0.057 (0.012)	-0.019 (0.016)
Church Attendance	1.057 (0.191)	-0.171 (0.043)	-0.006 (0.010)	0.079 (0.014)
Mode: Online	0.081 (0.162)	0.180 (0.043)	0.069 (0.010)	-0.026 (0.014)
Wordsum Score	0.931 (0.275)	0.401 (0.079)	0.103 (0.019)	0.017 (0.025)
Word Count (log)	-1.016 (0.802)	0.659 (0.220)	0.009 (0.053)	-0.089 (0.070)
Constant	-3.972 (0.586)	0.017 (0.165)	0.228 (0.039)	0.365 (0.052)
Observations	3,486	3,039	3,037	3,038
R ²		0.065	0.147	0.043
Log Likelihood	-1,084.839			