

## In Defense of Genopolitics

JAMES H. FOWLER *University of California, San Diego*

CHRISTOPHER T. DAWES *New York University*

**T**he *American Political Science Review* recently published a critique of an article we published in the *Journal of Politics* in 2008. In that article we showed that variants of the genes 5HTT and MAOA were significantly associated with voter turnout in a sample of 2,300 subjects from the National Longitudinal Study of Adolescent Health. Here, we address the critique first by conducting a replication study using an independent sample of 9,300 subjects. This study replicates the gene-environment interaction of the 5HTT gene variant with church attendance, but not the association with MAOA. We then focus on the general argument of the critique, showing that many of its characterizations of the literature in genetics and in political science are misleading or incorrect. We conclude by illustrating the ways in which genopolitics has already made a lasting contribution to the field of political science and by offering guidelines for future studies in genopolitics that are based on state-of-the-art recommendations from the field of behavior genetics.

**I**n 2005, the *American Political Science Review* published an article that used data on political attitudes from identical and fraternal twins in the United States and Australia to show that genes may play a role in political orientations (Alford, Funk, and Hibbing 2005). This was the first “twin study” published in political science, and a replication study was published shortly thereafter in *Behavior Genetics* (Hatemi et al. 2007). We followed up these studies of political attitudes with a twin study of political behavior. Using self-reported and validated turnout data from two sources, we showed that patterns of voter turnout and political participation were significantly more similar between identical twins than they were between fraternal twins (Fowler, Baker, and Dawes 2008).

These articles initiated a new field of inquiry that we call “genopolitics,” the study of the genetic basis of political behavior. And although the twin study design used in these early studies has been tested and replicated using alternative methods (Visscher et al. 2006; Yang et al. 2010), they are only a starting point that suggest *whether* and *how much* genes matter in general. Ultimately, however, scholars are interested in *how* genes matter. One way to address this question is to identify specific genes that are associated with a particular trait and so better understand the mechanism by which genes have their effect.

In 2008, we reported the results of the first candidate gene association (CGA) study of political behavior in an article published in the *Journal of Politics* (Fowler and Dawes 2008). In that study, we used data from the National Longitudinal Study of Adolescent Health

(Add Health) to show that voter turnout is significantly associated with variants of the 5HTT and MAOA genes. This was the first study to identify specific genes associated with a political behavior, and it was also the first political science study to test a gene-environment interaction. In a recent issue of the *American Political Science Review*, Charney and English (2012) (hereafter “CE”) published a general critique of genopolitics, focusing in particular on our 2008 candidate gene study. CE replicated our results from the earlier study and then suggested several alternatively specified models using the same data that we employed.

Here, we offer a reply to this critique. First, we describe candidate gene association tests and their limitations, reemphasizing our initial statement that our findings are only suggestive until they are replicated using independent samples. We then go on to conduct such a replication of our earlier work using new data from Wave IV of Add Health. The results show that the association between turnout and an interaction between 5HTT and church attendance replicates; however, the association with MAOA does not. We then conduct several tests with alternative specifications, showing that the 5HTT result is robust.

Turning to the general critique of genopolitics, we draw on literatures in both political science and genetics to show that many of the critique’s arguments are misleading or even incorrect. We then review the potential advantages of genopolitics research and prescribe guidelines for political science journals that are considering whether to publish candidate gene association studies. We conclude on an optimistic note, showing how genopolitics has stimulated an exchange of ideas between biology and politics.

At the outset, we would like to highlight several important points that we discuss later in greater detail. First, many of the criticisms raised by CE were already clearly acknowledged and explained in our original paper. Second, many of the points raised by CE are not specific to candidate gene studies, but are broader critiques of the choices typically made by political scientists doing empirical analysis. Third, we acknowledge that since our study first appeared there has been a

James Fowler is Professor of Medical Genetics and Political Science at the University of California, San Diego. (jhfowler@ucsd.edu).

Christopher T. Dawes is Assistant Professor of Politics at the Department of Politics, New York University.

We would like to thank Kevin Arceneaux, Robert Bond, Jan-Emmanuel de Neve, Pete Hatemi, John Hibbing, Peter Loewen, Rose McDermott, Alexandre Morin Chassé, Kevin Smith, and participants in the Third Annual Human Nature Group Annual Retreat for helpful comments on earlier drafts of this article.

growing concern over widespread failed replications of CGA studies (Chabris et al. 2012). These failed replications are likely due to the polygenic architecture of complex political traits, which means there are many genes of small effect that require large samples and/or independent replications to achieve adequate power. In addition, the problem of false-positive results may be more pervasive among candidate gene-environment interaction studies due to the added challenge of determining and measuring environmental moderators (Duncan and Keller 2011). Fourth, unlike CE, we believe that genetic studies, if they are done carefully, can be useful in helping gain a better understanding of political behaviors. Finally, we do not share CE's vision of scientific inquiry. The etiology, genetic or environmental, of any complex behavior is surely complicated, but we do not think most scientists would suggest that this complexity means we should give up this exploration. We advance our understanding of complex systems by studying their parts, with the goal of integrating this knowledge after enough is understood about the basic processes that lie at the heart of any phenomenon.

## CANDIDATE GENE ASSOCIATION STUDIES AND THEIR LIMITATIONS

Fowler and Dawes (2008) reported the results of a candidate gene association (CGA) study; it focused on 5HTT and MAOA as candidates because of their role in the serotonin system (Fowler and Dawes 2008, 579):

These two genes transcribe neurochemicals that exert a strong influence on the serotonin system in parts of the brain that regulate fear, trust, and social interaction (Bertolino et al. 2005; Eisenberger et al. 2007; Hariri et al. 2002; Hariri et al. 2005; Heinz et al. 2005; Meyer-Lindenberg et al. 2006). MAOA and 5HTT have been studied for more than 20 years, and much is known about the way different versions of their genes regulate transcription, metabolism, and signal transfers between neurons, all of which have an effect on social interactions (Craig 2007). In particular, the less transcriptionally efficient alleles of these genes have been associated with a variety of antisocial behaviors (Rhee and Waldman 2002).

Given that voting is considered a prosocial activity (Blais and Labbé-St-Vincent 2011; Dawes, Loewen, and Fowler 2011; Edlin, Gelman, and Kaplan 2007; Fowler 2006a; Gerber et al. 2008; Jankowski 2004; Knack 1992), we hypothesized that these genes influence voting. But we were also interested in the possibility that these genes might not act in isolation: “[A]n association between either MAOA and 5HTT and voting may not be direct. Instead, an association between a gene and turnout may be moderated by environmental factors” (Fowler and Dawes 2008, 583).

Although it is possible that there might have been other environmental moderators, we focused on church attendance because of its well-known relationship with voting and because it might have a specific effect on those who prefer to avoid social conflict:

Religious group activity in particular has been singled out as one of the strongest predictors of voter turnout, even more so than socioeconomic status (Olsen 1972, Sallach, Babchuk & Booth 1972). . . . Cassel (1999) suggests that the main reason for the association is that religious groups build a sense of belonging to a larger community. However, it may not be possible to build such a sense in people who are too averse to social conflict, since they will resist appeals to become involved. We therefore hypothesize that MAOA and 5HTT, when interacted with religious group activity, may be significantly associated with turnout. (Fowler and Dawes 2008, 583)

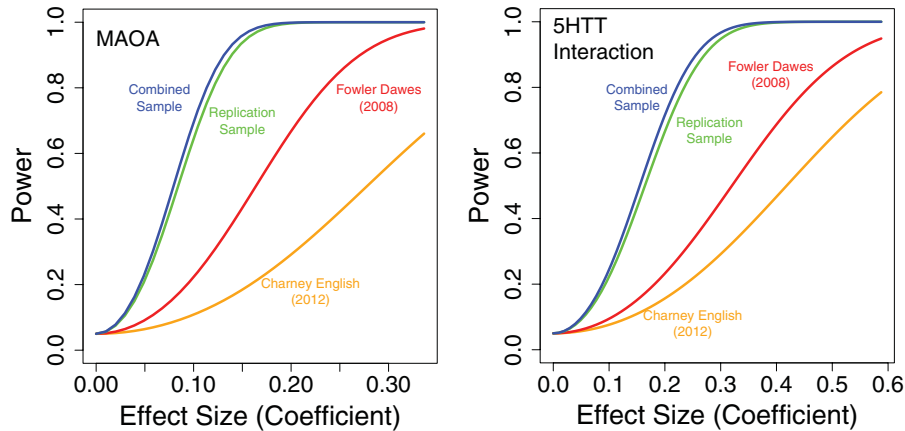
A CGA study tests whether a particular variant of a gene (an “allele”) is found more frequently than can be attributed to chance in a group exhibiting a particular trait compared to those without the trait. In our case, we asked whether the frequency of a particular allele is higher among voters than nonvoters, after noting several limitations of this method:

[A] significant association can mean one of three things: (1) The allele itself influences voting behavior; (2) the allele is in “linkage disequilibrium” with an allele at another locus that influences voting; or (3) the observed association is a false positive signal due to population stratification. (Fowler and Dawes 2008, 584)

“Linkage disequilibrium” refers to the fact that alleles near one another on a strand of DNA may be correlated with one another. Candidate gene studies typically cannot be used to determine whether the causal allele is the one identified or the one in linkage disequilibrium nearby. “Population stratification” occurs when groups have different allele frequencies due to their genetic ancestry. Through the process of natural selection, assortative mating, recombination, ecological adaptation, or genetic drift these groups may develop different frequencies of a particular allele. At the same time, the two groups may also develop divergent behaviors that are influenced not by the allele but completely by the environment in which they live. Once these two groups mix in a larger population, simply comparing the frequency of the allele to the observed behavior would lead to a spurious association.

In case-control CGA studies, researchers control for the problem of population stratification by including the race/ethnicity of the subject in the model or by analyzing data from each group separately. But this method does not guard against population stratification *within* each race or ethnic group, which has been shown to occur even in populations that are thought to be highly homogeneous (Price et al. 2009). Moreover, dividing the sample by group can dramatically reduce the study's power (more on this later). Conversely, family-based CGA studies eliminate the problem of population stratification by using family members, such as parents or siblings, as controls for ancestry. However, a major limitation of family-based studies is that they tend to be even more underpowered (Xu and Shete 2006).

Recent work has provided direct evidence of the long-held assertion that complex traits such as

**FIGURE 1. Power to Detect Genetic Association**

Relationship between effect size and power for the sample sizes we previously reported ( $N = 2,273$ ), those reported by CE ( $N = 803$ ), the independent Add Health replication sample we use in this article ( $N = 8,744$ ), and the combined (original and replication) sample ( $N = 9,821$ ). Left panel shows results for MAOA models, and right panel shows results for 5HTT interaction models. These results show that the replication and combined samples have substantially more power than the original tests. They also show that models tested by CE have very low power, suggesting a high probability of falsely confirming the null hypothesis.

intelligence (Davies et al. 2011) and personality (Vinkhuyzen et al. 2012) exhibit a polygenic architecture, with the heritable variation explained by many genetic variants with small effects. Benjamin et al. (2012b) showed that this is also the case for political attitudes. Combined, these results suggest that, to achieve the necessary power to detect small effects and thus reduce the risk of false-positive results, large samples are necessary. It also suggests that some previous candidate gene studies reporting large effect sizes may have been false positives due to a lack of power. Chabris et al. (2012) were unable to replicate several previously published associations for general intelligence even though they had adequate power to do so.

In Figure 1 we show the relationship between effect size and power for achieving significance at conventional levels ( $p = 0.05$ ) in candidate gene association studies. The power of a test is the probability of rejecting the null hypothesis when it is false. The red lines show the power to detect effects in a sample of the same size as that reported in Fowler and Dawes (2008), simulated by assuming the same distribution of the dependent variable, independent variables, and their interaction. The yellow lines show the power to detect effects in samples the same size as those reported by CE. Notice that CE's models are relatively underpowered. In contrast, the replication sample we describe later and the combined sample have much greater power. In fact, Figure 1 shows that this new sample is adequately powered to detect effects as small as 0.1 in the MAOA model and 0.2 in the 5HTT interaction model. The main reason for the difference in thresholds is that interaction tests are underpowered relative to tests for main effects (Esarey and Lawrence 2012). What this figure makes clear is that large sample sizes, replication, and meta-analyses are probably necessary to detect the small

effects we are seeking, especially in gene-environment interaction models (Duncan and Keller 2011).

## REPLICATION

In our previous work, we wrote that “[a]ssociation studies like ours require further replication before their findings can be truly considered anything more than suggestive; therefore more work needs to be done in order to verify and better understand the specific associations we have identified” (Fowler and Dawes 2008, 590). Fortunately, we recently gained access to data that allow us to conduct a replication of our 2008 findings. In 2012, Add Health released a set of genotype information for an expanded number of subjects in Wave IV who had not previously been genotyped in Wave III; see Fowler and Dawes (2008) or <http://www.cpc.unc.edu/projects/addhealth/> for more details about the Add Health study. This new data included 2,297 individuals who were previously genotyped for MAOA and 9,311 who were not, as well as 2,314 individuals who were previously genotyped for 5HTT and 9,410 individuals who were not. Restricting our analysis to those who were genotyped in Wave IV but not in Wave III gave us the opportunity to conduct an out-of-sample test of the significant associations previously reported between these genotypes and voting behavior. Table 1 presents summary statistics for the combined sample. Note that the incidence of genotypes is very similar between the smaller Wave III sample and the larger Wave IV sample; the correlation in genotype for those who were genotyped twice is 0.98 for MAOA and 0.99 for 5HTT (see Smolen et al. 2012, for full details on genotyping the new sample). All other variables were measured exactly as reported in Fowler and Dawes (2008), except where noted later.

**TABLE 1. Summary Statistics**

	Mean	SD	Min	Max	N
Vote in 2000 Election	0.46	0.50	0	1	13,668
Vote Frequency (Wave IV)	2.40	1.16	1	4	11,859
Long 5HTT (Wave III)	1.16	0.71	0	2	2,314
Long 5HTT (Wave IV)	1.16	0.72	0	2	11,454
High MAOA (Wave III)	1.19	0.86	0	2	2,297
High MAOA (Wave IV)	1.20	0.84	0	2	11,348
5HTT Triallelic genotype	0.96	0.72	0	2	11,344
Church Attendance	2.02	0.74	1	3	13,727
Male	0.47	0.50	0	1	13,829
Age	21.93	1.77	18	27	13,829
Age 18 to 20 in Nov. 2000	0.24	0.43	0	1	13,829
Age 20 to 22 in Nov. 2000	0.35	0.48	0	1	13,829
Age 22+ in Nov. 2000	0.40	0.49	0	1	13,829
White	0.67	0.47	0	1	13,829
Hispanic	0.14	0.34	0	1	13,829
Black	0.23	0.42	0	1	13,829
Asian	0.07	0.25	0	1	13,829
Native American	0.03	0.17	0	1	13,829

Note: All summary statistics are based on the combined sample except those denoting Wave III or IV.

**TABLE 2. Association of 5HTT and MAOA with Voter Turnout in a New Sample**

	All Subjects in the New Sample			Whites Only in the New Sample			All Subjects in the New Sample		
	Coef	SE	p	Coef	SE	p	Coef	SE	p
<b>Long * Attendance</b>	<b>0.19</b>	<b>0.07</b>	<b>0.01</b>	<b>0.17</b>	<b>0.09</b>	<b>0.05</b>			
Long 5HTT (LL,Ls)	-0.35	0.16	0.03	-0.36	0.19	0.05			
High MAOA							-0.01	0.04	0.86
Church Attendance	0.36	0.07	0.00	0.35	0.08	0.00			
Black	0.22	0.05	0.00				0.34	0.05	0.00
Hispanic	-0.15	0.07	0.03				-0.14	0.07	0.03
Asian	-0.29	0.09	0.00				-0.24	0.09	0.01
Native American	-0.36	0.14	0.01				-0.34	0.14	0.02
Age	0.10	0.01	0.00	0.10	0.01	0.00	0.09	0.01	0.00
Male	-0.10	0.04	0.02	-0.02	0.05	0.68	-0.17	0.04	0.00
Constant	-3.06	0.31	0.00	-3.04	0.37	0.00	-2.07	0.27	0.00
AIC	12391			8387			12601		
Deviance	12369			8373			12583		
Null deviance									
N families	8901			6071			8852		
N individuals	9266			6247			9213		

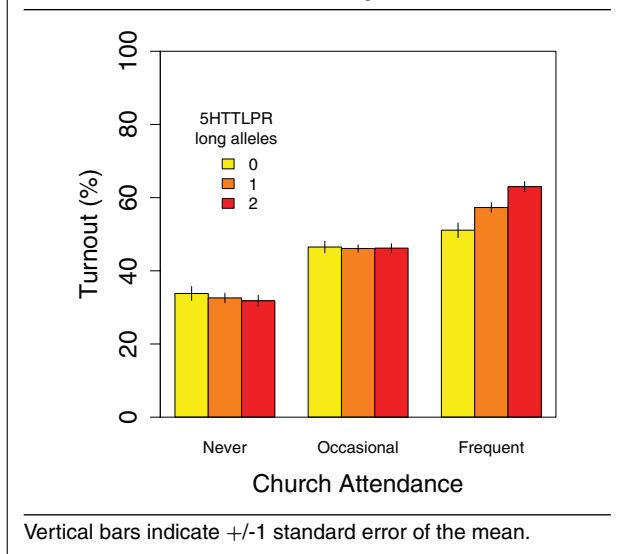
Note: Mixed-effects logistic regression models of voter turnout (1 = voted in 2000 election) using newly released subjects from Wave IV of the Add Health data not analyzed in Fowler and Dawes (2008). Cohen's d for Long \* Attendance is 0.10 in model 1 and 0.09 in model 2.

Using the same mixed-effects model reported in Fowler and Dawes (2008) on individuals who were not previously genotyped, Table 2 shows that the association between self-reported voting and the interaction of 5HTT and church attendance remained significant in the new sample (that is, excluding all individuals who were in the previously reported model). Church attendees with at least one “long” allele were 21% (SE 7%) more likely to vote than those with two “short” alleles. Moreover, the association replicated in a model

that included only white subjects. Figure 2 shows that this interaction is apparent in the raw data as well.

In contrast, Table 2 also shows that the association with MAOA did not replicate, and contrary to the originally reported association, it even produced a slightly negative coefficient. This result is unlikely to be due to low power – as mentioned earlier, Figure 1 shows that we have a sample size sufficient to detect effects as small as 0.1. We therefore suspect that the original MAOA result was a false positive.

**FIGURE 2. Distribution of Voter Turnout by 5HTTLPR and Church Attendance Based on Combined Add Health Sample Data**



Although we believe it would be sufficient to stop here and simply report the results in Table 2, in the remainder of this section on replication we address CE's methodological concerns with additional models that test the robustness of this replication and the overall relationship among 5HTT, church attendance, and voting.

**Dealing with Within-Family Correlation.** CE question our use of mixed-effect models to cope with multiple observations from the same family. The key focus of this critique is that “‘families’ in the Add Health dataset segregate into different ‘types’ whose members differ systematically” (9) and that this likely biases our result. Instead, they propose randomly choosing one sibling from each family to create an independent sample.

Although their recommendation is likely far too conservative (mixed-effect models are used throughout the social sciences), in the interest of testing whether our replication is sensitive to the choice of modeling framework we created a randomly chosen, fully independent sample with 8,744 unrelated individuals (in other words, we took only one subject from each family) and used a generalized linear model to estimate the association between 5HTT and the Add Health measure of self-reported turnout in the 2000 general election. Table S1 (Tables S1–S7 are in the Online Appendix) shows that this more conservative approach also yielded a significant association.<sup>1</sup> Importantly, none of the 8,744 subjects in this model were included in the original study because we did not have genotype information for them (they were genotyped in Wave IV but not in Wave III). Moreover, none came from households with oversampled twins or siblings, suggesting that CE's conjecture that oversampling would bias the results is incorrect.

<sup>1</sup> The Online Appendix can be found at <http://www.journals.cambridge.org/psr2013006>.

In Table S2, we combined the previously genotyped and newly genotyped samples to generate an estimate across both samples; however we continued to use only one individual per family ( $N = 9,821$ ). The results show that the estimated effect increased slightly, and the  $p$  value on the association was  $5 \times 10^{-3}$ . Note that the 95% confidence interval for the beta coefficient on the 5HTT interaction in the original study was [0.13 to 0.79] and in the replication study it was [0.06 to 0.32]. Although this suggests that the original sample was underpowered to detect an effect of the size estimated in the replication sample ( $0.45 > 0.32$ ), it does not invalidate the replication: In both samples, the association was positive and significantly different from zero, and a test of the difference in the coefficients produced by both samples was not quite significant ( $p = 0.07$ ). Given the much larger sample size in the replication study, however, we suspect that the true coefficient is closer to that we reported in the replication, which is confirmed by the model reporting results on the combined independent sample, suggesting a confidence interval of [0.07 to 0.33].

**Genotype Specification.** CE assert that the results we originally presented are sensitive to specification of the 5HTT variable. They claim that rather than coding the genotype as a 1 if a person had *at least* one long allele (combining the LL and Ls genotypes into one group), we should have coded it as 1 if a person had two long alleles (including only those with the LL genotype). In Table S3 we present results for alternative genotype coding criteria by including separate variables for both the LL and Ls genotypes. We found that the effect size of the Ls genotype is exactly half that of the LL genotype, suggesting an additive process that increases with the number of L alleles. In fact, an additive model of 5HTT produces a better fit (as evidenced by a lower AIC) than a model that includes the Ls and LL genotypes separately (and both models have a lower AIC than a fully saturated model that omits the genotype variables).

**Splitting the Sample.** Throughout their article CE argue against our findings by analyzing subsets of the overall sample and demonstrating a lack of statistical significance. However, it is important to consider that the failure to maintain a significant association found based on the total sample may simply be the result of a dramatic reduction in sample size. More concretely, any random culling of the data by  $X\%$  will increase the standard error by a factor of approximately  $\sqrt{100/(100 - X)}$  and, as noted earlier and demonstrated in Figure 1, will cause a large reduction in power. A more appropriate way to determine whether pooling by subsample is appropriate is to *test* whether the effect size varies by group. An easy way to do this is to interact the effect size with an indicator variable for the subsample in question.

In their first argument that the sample should be split, CE assert that individuals voting in their first election (ages 18–20) or their second election (ages 20–22) should not be pooled with older individuals. We tested the assertion that the association varies by age group by

interacting it with the 5HTT-Attendance interaction. Table S4 shows that we could reject the hypothesis that the interaction effect is different for 18–20 year olds ( $p = 0.17$ ) and for 20–22 year olds ( $p = 0.21$ ).

CE also assert that nonwhite individuals should not be pooled with whites because the effect for the whole population might be driven by an effect that only exists for a particular subgroup. We tested the assertion that the effect size varies by race or ethnicity by interacting it with the 5HTT-Attendance interaction. Table S5 shows that we could reject the hypothesis that the interaction effect is different for blacks ( $p = 0.49$ ), Hispanics ( $p = 0.72$ ), and Asians ( $p = 0.45$ ). The interaction with Native Americans is significant ( $p = 0.02$ ), but is positive, and the main interaction continues to be strongly significant ( $p = 2 \times 10^{-3}$ ), suggesting that Native Americans are not driving the overall association.

**Alternative Markers for 5HTT.** In addition to studying the original 5HTT measure, we used a newly available classification from the recently released Add Health data. In the previous release, 5HTT was classified according to whether individuals had a “short” or a “long” version of the 5HTTPLR allele in the promoter region. In the most recent release, a single nucleotide polymorphism (rs25531) was genotyped that has been shown to interact with 5HTTPLR to affect gene expression levels (Hu et al. 2006). Specifically, only those with the “long” genotype and adenine (A) at rs25531 evinced higher expression of 5HTT. We used this “tri-allelic” genotype to test the association among 5HTT, church attendance, and voting. In Table S6 we show that this alternative classification also yielded a significant interaction with church attendance ( $p = 1 \times 10^{-3}$ ).

**Alternative Phenotypes.** CE assert that voting in a single election is not sufficient to measure voting behavior. We have more to say about the broader issues related to this critique in the next section, but we were able to take advantage of new data to test CE’s assertion. Although Add Health does not have additional measures covering voting behavior in other elections, Wave IV does contain a question tapping one’s overall propensity to vote: “How often do you usually vote in local or statewide elections?” Four responses were permitted for this question: never, sometimes, often, and always. This question is not strictly comparable to the question we previously used as a measure of voting because the Wave III question is about a national election held in 2000 and the Wave IV question is about state and local elections held through 2008. Nonetheless, the polychoric correlation between these two measured variables was positive and highly significant (0.55, 95% CI 0.53 to 0.57).

To test whether the association with 5HTT is sensitive to phenotype specification, we regressed the Wave IV voting measure on the same independent variables shown in our other models. Table S7 shows that the interaction between the genotype and church attendance remained significant ( $p = 0.02$ ). This model assumes that the categorical variable is continuous, so to be sure this assumption is not driving the result, we also

estimated an ordered logit specification. The interaction between the genotype and church attendance was also significant in this alternative model ( $p = 0.03$ ).

There are important limitations to note regarding our replication exercise. Although our analysis is based on a much larger sample than previously employed and is therefore more likely to be adequately powered, the previous track record of candidate gene-environment interaction studies suggests that our results should be evaluated cautiously. In addition, we still run the risk of finding false positives due to population stratification. Even though we compared estimates across groups and restricted our analysis to whites, we lacked additional genetic information to control for unobserved population structure. This information could be provided using genome-wide association study (GWAS) samples once they become available. Another alternative is to conduct a replication based on a non-U.S. sample in which the potential confounding relationship between allele frequency and the environmental determinants of voting behavior is unlikely to follow the same pattern. Finally, it should be pointed out that a previously reported association between depression and an interaction of 5HTT and stressful life has not been reproduced by follow-up studies (Duncan and Keller 2011; Risch et al. 2009). Although we conducted a *direct* replication (Duncan and Keller 2011) of a relationship among 5HTT, church attendance, and voter turnout, the hypothesis that the effect of 5HTT is moderated by stressful events has clearly been undermined by failed replications.

## BROADER ISSUES IN THE STUDY OF VOTER BEHAVIOR

CE make several arguments that suggest it is impossible to measure voter behavior and that we either inadequately defined our measures or we were doing something that was inconsistent with prior literature. In this section we address these critiques with the goal of documenting that the choices we made and the limitations we originally highlighted are consistent with standard practices in the discipline.

For example, CE argue that “FD distinguish their cases—‘voters’—from controls, ‘nonvoters,’ on the basis of their ‘voting behavior,’ but do not define these terms” (3). Yet we clearly stated that our measure of voting behavior is the answer to this question: “Did you vote in the most recent [2000] presidential election?” (Fowler and Dawes 2008, 584). In fact, CE themselves even reference our definition when they assert, “It is apparent that responses to the question, ‘Did you vote in the most recent presidential election?,’ do not provide information concerning what is usually intended by the expression ‘voting behavior’”(3). They also complain that “FD provide no further specification of this new phenotype” (4) and call it “underspecified” (5), suggesting that we were somehow doing things differently than other scholars had in the past.

Their assertions mischaracterize the literature on voting. Self-reported turnout has been used as a mea-

sure of voting behavior in dozens (perhaps hundreds) of published studies (see, for example, Bowler and Donovan 2002; Goldstein and Freedman 2002; Jackson and Carsey 2007; Kan and Yang 2001; Koch 1998; Timpone 1998). In fact, the National Election Study has asked this question at each election since 1948 precisely with the hope that it would be used in research on voting behavior.

Relatedly, CE complain that we “dismiss” the problem of *overreporting*: the tendency for some survey respondents who did not vote to say that they did (5). However, we were not dismissive. In the 2008 article, we wrote,

[I]t would be preferable to have information about validated turnout because of the well-known problem of over-reporting—many people who say they voted actually did not (Karp and Brockington 2005). However, Fowler, Baker, and Dawes (2008) show that a substantial genetic component exists for both validated and self-reported turnout, and they do not find a statistically meaningful difference in the size of the component for the different measures. (Fowler and Dawes 2008, 584)

In fact, in other work we have often relied on validated data as an additional measure of voting behavior (e.g., Bond et al. 2012; Fowler 2005; 2006b). However, we also recognize that validated turnout data are not a panacea. A recent comparison of the two measures concluded, “Using government records in lieu of self-reports, which can be both time-consuming and expensive, appears to inject more error than accuracy into measurements of registration and turnout” (Berent, Krosnick, and Lupia 2011, 72) Thus, we have usually sought to do studies using both validated and self-report measures when they are available (see Fowler, Baker, and Dawes 2008, for an example).

Noting that “inherent imprecision is a natural feature of human language” (5), CE point out that voting is affected by context—in countries where voting is mandatory, dangerous, or irrelevant, people may behave differently than they do in the United States. Yet they draw from this feature the inference that voting is “a different behavior” (5) in each of these circumstances. The goal of CE seems to be to make measurement impossible by persuading the reader that all acts of voting are unique and that we cannot even be sure that we know what it means when someone says “I voted” in a specific context.

Although it is true that semantics makes quantitative evaluation of behavior difficult, systematic application of well-described measures can help us overcome the problem of imprecise language. In our case, we relied on a replicable measure that has been used for several decades. In fact, it is so replicable that CE themselves were able to reproduce our original analysis. We therefore can imagine other scholars testing our hypotheses in different contexts using the same instrument. If other studies produce different results, one possible explanation is that context matters.

CE conclude their argument: “As commonly intended, however, the expression ‘voting behavior’ refers to a quantitative variable (i.e., one votes *more or*

*less frequently*)” (5). In this statement, not only do CE misconstrue the literature on voting but they also mischaracterize what is a quantitative variable. A yes/no measure is still a “quantity” measured by 1 and 0, and it can be used to infer a latent *continuous* variable that indicates the probability of taking a certain action. In fact, this is exactly the assumption underlying the replication models they conducted. In logistic regression, the assumption is that there is an unobserved value  $p$  that indicates the probability of an event (in this case, a vote) that is a function of several independent variables, and the goal is to fit parameters that define this  $p$  such that it maximizes the likelihood of the observed distribution of events (e.g., vote choices; King 1998).

## BROADER ISSUES IN GENETICS

CE make several key points in their discussion of broader issues in genetics that we believe should be directly addressed because they are misleading or even incorrect.

**Findings from Medical Genetics.** CE argue that “the search for genes that could predict prevalent and devastating behavioral phenotypes such as schizophrenia and autism, not to mention global killers such as diabetes and hypertension, has to date been unsuccessful” (2). They go on to identify just three single-nucleotide polymorphisms (SNPs) as “the few reliably reproduced associations between an [sic] SNP and a given phenotype of such a magnitude that the SNP can be called a risk factor” (13), but this argument is highly misleading. For example, a recent review of the literature shows that since 2007, 40 genetic loci have been discovered for type I diabetes and 50 have been discovered for type II diabetes (Visscher et al. 2012). These loci are among 384 SNPs associated with autoimmune and metabolic diseases with great enough significance in large samples that they are considered to be reliable associations. In fact, enough progress has been made that “it should be possible to identify groups of individuals who are at a substantially greater-than-average risk for diabetes” (Visscher et al. 2012, 17). Genetic susceptibility scores have already been created and are being applied for other complex phenotypes such as body mass index (Speliotes et al. 2010). We do not claim that medical genetics has a complete understanding of complex diseases or that prediction at the individual level is a viable reality yet; however, neither is it accurate to summarily dismiss what has been accomplished so far.

**The Current “Paradigm” in Genetics.** CE argue that “CGA studies of complex behavioral traits (such as political behaviors) rely on an outdated genetic paradigm” (2). However, the journal *Behavior Genetics* recently updated its guidelines on the publication of candidate gene studies (Hewitt 2012) to reflect the latest research in genetics, and it continues to publish CGA studies. Moreover, these studies are still viewed as beneficial, even by critics. The authors of a recent highly critical review of candidate gene studies write that they “may well prove to be important or even



central for understanding the etiology of psychiatric disorders. At issue is how to separate the wheat from the chaff” (Duncan and Keller 2011,1047–48) Thus the “paradigm” is not at issue—it is the methodology that proves challenging. The solution is not to abandon candidate gene studies, but rather to implement guidelines for their publication that reduce the likelihood of false positives.

**Gene Transcription and Stability.** It is important to understand that what takes place at the molecular level is complex. However, CE’s description of this complexity is highly misleading. For example, they write, “Once considered the paragon of stability, DNA is subject to all manner of transformation. For example, retrotransposons or “jumping genes” comprise 45% of the human genome, move about the genome by a copy-and-paste mechanism changing DNA content and structure” (10–11). Yet consider that the rate of mutation per generation per site in human DNA is estimated to be about 1 in 100 million (Lynch 2010). This means that all but a few dozen of the 3 billion base pairs in an individual’s DNA will be exactly the same *throughout our reproductive lifetimes*. Thus, by and large, the genes we are born with are the genes we will die with. It is true that the environment plays a critical role in gene expression, myelination, and other biological processes that influence the way genes affect us, but it is extremely disingenuous to argue that these measures of our biological inheritance are unstable.

CE further assert, “Genes do not regulate the extent to which they are capable of being transcribed in any obvious, unidirectional manner” (10). This claim is directly contradicted by evidence from more than 5,000 genes that shows that transcription explains nearly 40% of the variation in levels of expression in mammals (Schwanhäusser et al. 2011). Although this evidence also suggests an important role for the environment, it shows that heritable DNA sequence variation does have an important impact on expression. In particular, a recent review of the literature on the 5HTTLPR polymorphism shows that the transcriptional efficiency of the “long” version of this genotype is associated with significantly higher expression than the shorter allele in both mice and humans (Murphy and Lesch 2008). Causal evidence of the gene’s effect on expression is especially strong in mouse studies, because many of them “knock out” (delete) or modify the gene in question (Murphy and Lesch 2008).

This does not mean that gene-gene and gene-environment effects on transcription are absent, and recent research is exploring a variety of factors that modify gene expression (Cheung and Spielman 2009), including cell type (Dimas et al. 2009). However, it is incorrect to imply that it is impossible to detect a relationship between genotypes and transcription levels. Given that we do not know what many genes do, the best place to start is to study genotypes that we know have a measurable effect on the molecules they transcribe.

**Pleiotropy.** CE list many phenotypes that have been associated with variation in 5HTT and MAOA and

ask, “How is it possible that the same polymorphisms of the same gene could simultaneously predict (or be risk factors for) so many different phenotypes?” (13). Genes playing a role in important hormonal systems that regulate social behavior are likely to influence a wide variety of outcomes, a phenomenon known as “pleiotropy” (Lande 1980). Depression and personality factors are two possible mediating mechanisms, but there may be many others.

CE question the idea of an underlying factor that may cause several of the phenotypic outcomes (an “endophenotype”). However, they neglect to point out that systems such as the serotonin system influence a wide variety of behaviors via different mechanisms (Murphy and Lesch 2008). This means that a single genotype may influence several endophenotypes, each of which is associated with particular phenotypes; there is strong evidence that this is true for 5HTT. Knock-out studies in mice where the 5HTT gene is completely removed show that a large number of phenotypes are affected, and the results are so well accepted that pleiotropy in 5HTT has been proposed as a reason for evolutionary conservation of diversity in this gene (Murphy and Lesch 2008).

Consider a genotype we have already been studying for decades: sex. A wide range of political phenomena are influenced by a person’s sex (Hatemi et al. 2012), and although we do not claim that these effects are all purely genetic, neither do we doubt a reported association between sex and one political behavior merely because it also predicts other political behaviors. Sex can influence specific hormonal responses, but it can also induce gender-specific roles in the self and others (Hatemi et al. 2012). Each of these could be considered an endophenotype with many phenotypic consequences.

Another important consideration is that many of the phenotypes listed by CE have not been replicated, so it may be premature to ask what such different phenotypes have in common with one another. For example, our replication efforts here suggest that voting should be removed from the list of those phenotypes associated with MAOA. At the same time, however, our confidence that 5HTTPLR plays a role is now increased by successful replication, so (for now) it belongs on the list.

CE continue: “Consider that the polymorphisms of MAOA- $\mu$ VNTR and 5-HTTLPR are the only polymorphisms for these two genes for which there are data in a large data set such as Add Health and that MAOA and 5-HTT are two out of the *eight genes in total* (emphasis in original) for which there are genotypic data. What is the likelihood that these same polymorphisms on these same genes will conveniently turn out to be the genetic key to so much human behavior?” (13). This question is misleading in two ways. First, we never called these genes “the key to human behavior.” Consider what we actually wrote:

It is important to emphasize that there is likely no single “voting gene”—the results presented here suggest that at least two genes do matter and there is some (likely



large) set of genes whose *expression, in combination with environmental factors, influences political participation.* (Fowler and Dawes 2008, 590)

Second, the investigators of the National Longitudinal Study of Adolescent Health did not choose their genes randomly—they focused on genes that had already shown promise in explaining social behavior and that were linked to behavioral and psychiatric outcomes. According to the study researchers,

Saliva samples were collected from full siblings or twins to genotype DNA for seven candidate polymorphisms. These candidate genes have been reported to be associated with individual differences in behavior related to mental health; are reported to be functional, exonic, in promoter regions, or affect gene expression; are expressed in the brain; and have *prima facie* involvement in neurotransmission. (Smolen and Hewitt 2003, 42–43)

It should thus not be surprising that biomarkers for the genes they chose have an effect on many social behaviors, and we would have a better chance of finding more than one association among them.

**Stochastic vs. Deterministic Outcomes.** CE write, “The problem is that a large set of genes (and if large enough, then we are simply talking about the human genome and hence the human organism), the transcriptional activity of which is influenced by the environment and each others’ functional products, is incompatible with the expectation that two genes could predict voter turnout” (30). This is a disingenuous critique. We clearly acknowledged that there are multiple genetic and environmental causal factors that underlie turnout: “There is some (likely large) set of genes whose expression, in combination with environmental factors, influences political participation” (Fowler and Dawes 2008, 590).

Our usage of the word “predict” in our original article was meant to convey that there is a nonrandom, systematic relationship between at least two (and probably very many more) genes and voter turnout. In this article we show that one of these relationships replicates and one does not. We therefore conclude that one of these genes “predicts” turnout, even though we in no way believe and have never argued that it is the only such gene. We were very careful in 2008 to make this clear: “[I]t is important to emphasize that there is likely no single ‘voting gene’” (Fowler and Dawes 2008, 590) In fact, we deliberately (emphasis added) chose to publish associations with two genes in one paper (rather than two separate papers) to emphasize this point.

CE go on to argue that “CGA studies depend on the assumption that the presence of a particular allele entails that it is turned on. . . . We can no longer assume, however, that the presence of a particular allele entails that it is capable of being transcribed in the manner associated with that allele, because it may be epigenetically silenced” (10). Again, this argument demonstrates a profound misunderstanding of scientific inference. Although it is true that gene expression is variable and it can be sensitive to the environment

and other genotypes, this does not mean that the gene plays no role in influencing a particular behavior. Nor does it preclude studies that take this stochasticity into account by testing the hypothesis with many observations and under many conditions.

**Polygenic Effects.** CE emphasize the multifactorial nature of genetic effects, writing that “most human traits with a genetic component are influenced by a vast number of genes of small effect” (11) and that “[r]esearchers who conduct CGA studies err when they suppose that multifactorial traits exhibit a genotype-phenotype relationship analogous to that of monogenic disorders” (12). They write this in such a way that it implies that we are not aware of these complexities. Yet, note what we wrote four years ago: “[T]he vast majority of phenotypes are ‘polygenic,’ meaning they are influenced by multiple genes (Mackay 2001; Plomin 2008) and are shaped by a multitude of environmental forces” (Fowler and Dawes 2008, 581).

CE conclude that the existence of polygenic effects “calls into question the underlying assumption of CGA studies” (11) and makes genetic studies “particularly problematic” (11). In stark contrast, we argued as follows:

[S]imple association models between genotype and phenotype are an important first step to establish candidate genes, but they are not the end of the story. It is also important to investigate the extent to which genetic associations are moderated by environmental factors (“environmental modifiers”) and other genes (“genetic modifiers”). (Fowler and Dawes 2008, 581)

At the heart of this disagreement is a fundamental difference in the way we and CE approach scientific inquiry. CE claim that because the system we are studying is complex, it is pointless to examine any of its parts. But this philosophy would preclude nearly all of social science. Social systems are inherently complex, and observational and experimental studies attempt to reduce complexity by isolating parts of these systems to determine which factors and processes are most important for a given outcome. The hope is that these factors can then be integrated into a more holistic theory at the system level.

To be clear, we are sympathetic to systems-level approaches. In other work, we advocate for a better understanding of interdependence and its effect on a number of political phenomena (e.g., Christakis and Fowler 2009). However, we would never make the argument that macro-level approaches are a substitute for micro-level research. The truth is that we need both.

## BENEFITS OF GENOPOLITICS RESEARCH

There are a number of interesting ways that genopolitics research may contribute to existing literatures in political science. One potentially beneficial application is to use genetic information as a control variable in standard nongenetic studies (Beauchamp et al. 2011).

At least part of the variation in political attitudes and behaviors has been demonstrated to be due to genetic variation. Standard models relegate this variation to the residual, and thus accounting for it may allow us to more precisely estimate the effects of other factors of interest.

Another possibility, already being explored by economists (Norton and Han 2008), is to use genetic variants as instrumental variables. The idea is to use the fact that genotypes are randomly passed from parents to offspring during meiosis as a type of natural experiment. In other words, genetic data could be used to construct an instrumental variable allowing causal inference to be based on observational data. However, this approach faces many technical challenges, including credibly meeting the exclusion restriction due to the possibility of pleiotropy (Benjamin et al. 2012b).

Yet another innovation that could potentially come out of genopolitics research is the ability to predict political traits using genetic data (Benjamin et al. 2012a). A major goal of medical genetics is to diagnose and treat diseases before they can otherwise be detected. However, Benjamin et al. (2012b) suggest that our ability to predict political behaviors and attitudes out of sample is presently limited and will ultimately require much larger samples.

In addition to these potential applications to existing literatures, genopolitics has already helped spur inquiries into a number of phenomena on the causal pathway from genes and environment to political outcomes. For example, the study of personality was recently revived by political scientists who wondered why genes might play a role in political behavior:

Five-factor models of personality trait structure have thrived in psychology for two decades, and rigorous attention to these models offers an obvious complement to recent developments in the exploration of links between politics and genetics. Our concern with this situation has motivated the present investigation. (Mondak et al. 2010, 20)

The study of neuropolitics has been similarly influenced: “The new science of human nature demands that we recognize that genes are the institutions of the human body. They regulate the neurological processes that drive social and political behavior” (Fowler and Schreiber 2008, 914). The new field of political physiology specifically seeks to understand the physiological mechanisms that genes influence: “Indeed, given that political and social attitudes are heritable and that amygdala activity also has been traced to genetics, genetic variation relevant to amygdala activity could affect both physiological responses to threat and political attitudes bearing on threats to the social order” (Oxley et al. 2008, 1669).

Thus, genetic studies of political behavior are not only beneficial in their own right. They may help integrate the social sciences with one another and may bring us closer to a “consilient” unification with the natural sciences (Wilson 1998) that will allow us to

better understand not only politics but also what it means to be human.

## RECOMMENDATIONS FOR FUTURE WORK

Many of the questions raised by CE regard matters of empirical analysis in general, rather than genetics in particular. Our 2008 candidate gene study faced the same kinds of challenges that nearly all political science studies face. However, our replication exercise speaks to an incredibly important concern that particularly affects genetic association studies—that of power.

As Benjamin et al. (2012b) recently demonstrated, political traits have a polygenic architecture, meaning that the heritable variation in these traits is explained by many genes with individually small effects. This implies that very large samples are necessary to reliably detect these effects and that underpowered CGA studies run the risk of producing false positives (Benjamin et al. 2012a). This power issue is even more critical when conducting a genome-wide study based on hundreds of thousands of genetic variants because of the need for required corrections for multiple testing.

The typical approach now in behavior genetics, as well as in most good science, is to search for *independent* sources of information by which to evaluate hypotheses. In fact, this is the new policy at the journal *Behavior Genetics*: “To avoid publishing findings that will not replicate, we recommend that authors conduct a direct replication analysis, prior to publication, such that the same predictor(s), outcome variable, and statistical model are tested in an independent sample” (Hewitt 2012, 1).

A recent critical review of candidate gene-environment interaction studies suggests that “direct” replication is the best way to avoid false positives (Duncan and Keller 2011). This means that the phenotype, the environmental modifier, and the genetic marker should all be measured in the same way. The concern is that authors might report “indirect” replications after they discovered that the direct replication did not produce a confirmation of the original result. We recommend that the *American Political Science Review* and other political science journals adopt the same policy as *Behavior Genetics* (Hewitt 2012). This will allow political scientists and geneticists to continue to develop the field of genopolitics while minimizing the likelihood of false discovery.

## CONCLUSION

The publication of Alford, Funk, and Hibbing’s (2005) seminal article in the *American Political Science Review* on the genetic basis of political attitudes came as something of a shock to many scholars in political science. At the time, the vast majority of the discipline had little exposure to the natural sciences. Meanwhile, many in the behavior genetics community wondered what all the fuss was about, because the original analyses of the same dataset had been published nearly two decades earlier (Martin et al. 1986). But political

scientists have been rapidly learning the new methods in genetics that have been developed in the past few years (Cranmer and Dawes 2012; Dawes and Fowler 2009; Hatemi, Funk, et al. 2009; Littvay 2012; Loewen and Dawes 2012; McDermott and Hatemi 2011; Settle et al. 2010; Settle, Dawes, and Fowler 2009; Smith and Hatemi 2012; Stam, Von Hagen-Jamar, and Worthington 2012; Verhulst 2012; Weber, Johnson, and Arceneaux 2011), and we are now working directly with geneticists on a variety of political outcomes and behaviors (Arceneaux, Johnson, and Maes 2012; Benjamin et al. 2012b; Eaves and Hatemi 2008; Fowler, Baker, and Dawes 2008; Hatemi et al. 2007; 2010; 2012; Hatemi, Alford, et al. 2009; Hatemi, Dawes, et al. 2011; Hatemi, Gillepsie, et al. 2011; Hatemi, Medland, and Eaves 2009; Klemmensen et al. 2011; McDermott et al. 2009; Medland and Hatemi 2008; Smith et al. 2012; Verhulst and Estabrook 2012; Verhulst, Hatemi, Eaves 2012a; 2012b; Verhulst, Hatemi, and Martin 2010). As a result, we have attained new literacy in methods including twin studies, candidate gene association studies, and genome-wide analyses that incorporate hundreds of thousands of genetic markers into our analysis.

It is important to note that we chose the title “In Defense of Genopolitics,” rather than “In Defense of Candidate Gene Studies.” We believe that genopolitics has the potential to make important contributions to the study of politics and also that CGA studies, like any other study, are open to criticism when not executed carefully. The CGA study was the dominant paradigm in medical genetics for a long time, but has recently been supplanted by genome-wide approaches both because of the rapidly falling cost of genotyping and their failure to be replicated (Chabris et al. 2012; Ioannidis 2005). This failure may be due to a variety of factors including differences in samples and measures, a lack of power, and a publication bias toward significant results. However, these issues can be addressed by designing studies that are adequately powered and/or draw on independent replication samples. The replication presented here illustrates this point; one of our original results held, whereas another one was likely a false positive.

It is also important to point out that, although CGA studies are hypothesis driven, they are constrained by an imperfect understanding of the pathways linking genes to behavior (Pearson and Manolio 2008). Therefore, scholars should be careful not to put too much weight on any given hypothesis based on a single finding. Significant associations are the beginning rather than the end of a long process of inquiry.

In conclusion, we think the future is bright for genopolitics research. Rather than taking an approach that assumes the effect of genes on behavior is hopelessly complicated and too hard to understand, we prefer to conduct open empirical inquiries that lay out the advantages and disadvantages of the methods we use to measure the world. We invite scholars to come forward with their own data and analyses, and together we can build up our theoretical and empirical understanding of the role biology plays in politics.

## References

- Alford, J. R., C. L. Funk, and J. R. Hibbing. 2005. “Are Political Orientations Genetically Transmitted?” *American Political Science Review* 99 (2): 153–67.
- Arceneaux, K., M. Johnson, and H. H. Maes. 2012. “The Genetic Basis of Political Sophistication.” *Twin Research and Human Genetics* 15 (1): 34–41.
- Beauchamp, J. P., D. Cesarini, M. Johannesson, M. J. H. M. van der Loos, P. D. Koellinger, P. J. F. Groenen, J. H. Fowler, J. N. Rosenquist, A. R. Thurik, and N. A. Christakis. 2011. “Molecular Genetics and Economics.” *Journal of Economic Perspectives* 25 (4): 1–27.
- Benjamin, D., et al. 2012a. “The Promises and Pitfalls of Genoeconomics.” *Annual Review of Economics* 4: 627–62.
- Benjamin, D., et al. 2012b. “The Genetic Architecture of Economic and Political Preferences.” *Proceedings of the National Academy of Sciences* 109 (21): 8026–31.
- Berent, M. K., J. A. Krosnick, and A. Lupia. 2011. “The Quality of Government Records and “Overestimation” of Registration and Turnout in Surveys: Lessons from the 2008 ANES Panel Study’s Registration and Turnout Validation Exercises.” Working Paper NES s012554. August 2011 version. Ann Arbor, MI: American National Election Studies.
- Bertolino, A., G. Arciero, V. Rubino, V. Latorre, M. De Candia, and V. Mazzola. 2005. “Variation of Human Amygdala Response during Threatening Stimuli as a Function of 5HTTLPR Genotype and Personality Style.” *Biological Psychiatry* 57: 1517–25.
- Blais, A., and S. Labbé-St-Vincent. 2011. “Personality Traits, Political Attitudes, and the Propensity to Vote.” *European Journal of Political Research* 50: 395–417.
- Bond, R. M., C. J. Fariss, J. J. Jones, A. D. I. Kramer, C. Marlow, J. E. Settle, and J. H. Fowler. 2012. “A 61-million-person Experiment in Social Influence and Political Mobilization.” *Nature* 489: 295–98.
- Bowler, S., and T. Donovan. 2002. “Democracy, Institutions and Attitudes about Citizen Influence on Government.” *British Journal of Political Science* 32: 371–90.
- Cassel, C. A. 1999. “Voluntary Associating, Churches, and Social Participation Theories of Turnout.” *Social Science Quarterly* 80 (3): 504–17.
- Chabris, C., et al. 2012. “Most Reported Genetic Associations with General Intelligence Are Probably False Positives.” *Psychological Science* 23 (11): 1314–23.
- Charney, E., and W. English. 2012. “Candidate Genes and Political Behavior.” *American Political Science Review* 106 (1): 1–34.
- Cheung, V. G., and R. S. Spielman. 2009. “Genetics of Human Gene Expression: Mapping DNA Variants That Influence Gene Expression.” *Nature Reviews Genetics* 10: 595–604.
- Christakis, N. A., and J. H. Fowler. 2009. *Connected: The Surprising Power of Our Social Networks and How They Shape Our Lives*. New York: Little Brown.
- Craig, I. W. 2007. “The Importance of Stress and Genetic Variation in Human Aggression.” *Bioessays* 29 (3): 227–36.
- Cranmer, Skyler J., and Christopher T. Dawes. 2012. “The Heritability of Foreign Policy Preferences.” *Twin Research and Human Genetics* 15 (1): 52–59.
- Davies, G., et al. 2011. “Genome-Wide Association Studies Establish that Human Intelligence is Highly Heritable and Polygenic.” *Molecular Psychiatry* 16: 996–1005.
- Dawes, C. T., and J. H. Fowler. 2009. “Partisanship, Voting, and the Dopamine D2 Receptor Gene.” *Journal of Politics* 71 (3): 1157–71.
- Dawes, C. T., P. J. Loewen, and J. H. Fowler. 2011. “Social Preferences and Political Participation.” *Journal of Politics* 72 (4): 1189–98.
- Dimas, A. S., et al. 2009. “Common Regulatory Variation Impacts Gene Expression in a Cell Type-dependent Manner.” *Science* 325: 1246–50.
- Duncan, L. E., and M. C. Keller. 2011. “A Critical Review of the First 10 Years of Candidate Gene-by-environment Interaction Research in Psychiatry.” *American Journal of Psychiatry* 168: 1041–49.
- Eaves, L. J., and P. K. Hatemi. 2008. “Transmission of Attitudes toward Abortion and Gay Rights: Parental Socialization or Parental Mate Selection?” *Behavior Genetics* 38: 247–56.
- Edlin, A., A. Gelman, and N. Kaplan. 2007. “Voting as a Rational Choice.” *Rationality and Society* 19 (3): 293–314.

- Eisenberger, N., B. Way, S. Taylor, W. Welch, and M. Lieberman. 2007. "Understanding Genetic Risk to Aggression: Clues from the Brain's Response to Social Exclusion." *Biological Psychiatry* 61: 1100–8.
- Esarey, J., and J. Lawrence. 2012. "Overconfidence in Testing of Interaction Hypotheses: Determining and Controlling the False Positive Rate." Unpublished manuscript
- Fowler, J. H. 2005. "Turnout in a Small World." In *The Social Logic of Politics: Family, Friends, Neighbors, and Workmates as Contexts for Political Behavior*, ed. A. Zuckerman. Philadelphia: Temple University Press, 269–87.
- Fowler, J. H. 2006a. "Altruism and Turnout." *Journal of Politics* 68 (3): 674–83.
- Fowler, J. H. 2006b. "Habitual Voting and Behavioral Turnout." *Journal of Politics* 68 (2): 335–44.
- Fowler, J. H., L. A. Baker, and C. T. Dawes. 2008. "Genetic Variation in Political Participation." *American Political Science Review* 102 (2): 233–48.
- Fowler, J. H., and C. T. Dawes. 2008. "Two Genes Predict Voter Turnout." *Journal of Politics* 70 (3): 579–94.
- Fowler, J. H., and D. Schreiber. 2008. "Biology, Politics, and the Emerging Science of Human Nature." *Science* 322 (5903): 912–14.
- Gerber, A. S., D. P. Green, and C. W. Larimer. 2008. "Social Pressure and Voter Turnout: Evidence from a Large-scale Field Experiment." *American Political Science Review* 102 (1): 33–48.
- Goldstein, K., and P. Freedman. 2002. "Campaign Advertising and Voter Turnout: New Evidence for a Stimulation Effect." *Journal of Politics* 64 (3): 721–40.
- Hariri, A. R., et al. 2005. "A Susceptibility Gene for Affective Disorder and the Response of the Human Amygdala." *Archives of General Psychiatry* 62: 146–52.
- Hariri, A. R., V. S. Matty, A. Tessitore, B. Kolachana, F. Fera, and D. Goldman. 2002. "Serotonin Transporter Genetic Variation and the Response of the Human Amygdala." *Science* 297: 400–3.
- Hatemi, P. K., J. R. Alford, J. R. Hibbing, N. G. Martin, and L. J. Eaves. 2009. "Is There a 'Party' in Your Genes?" *Political Research Quarterly* 62 (3): 584–600.
- Hatemi, P. K., C. T. Dawes, A. Frost-Keller, J. E. Settle, and B. Verhulst. 2011. "Integrating Social Science and Genetics: News from the Political Front." *Biodemography and Social Biology* 57: 67–87.
- Hatemi, P. K., C. Funk, S. E. Medland, H. H. Maes, N. G. Martin, and L. J. Eaves. 2009. "Genetic Influences on Social Attitudes over the Life Course." *Journal of Politics* 71 (3): 1141–56.
- Hatemi, P. K., Gillespie N. A., et al. 2011. "A Genome-wide Analysis of Liberal and Conservative Political Attitudes." *Journal of Politics* 73 (1): 271–285.
- Hatemi, P. K., Hibbing J. R., et al. 2010. "Not by Twins Alone: Using the Extended Twin Family Design to Investigate the Genetic Basis of Political Beliefs." *American Journal of Political Science* 54 (3): 798–814.
- Hatemi, P. K., R. McDermott, J. M. Bailey, and N. G. Martin. 2012. "The Different Effects of Gender and Sex on Vote Choice." *Political Research Quarterly* 65 (1): 76–92.
- Hatemi, P. K., S. E. Medland, and L. J. Eaves. 2009. "Genetic Sources for the Gender Gap?" *Journal of Politics* 71: 1–13.
- Hatemi, P. K., S. E. Medland, K. I. Morley, A. C. Heath, and N. G. Martin. 2007. "The Genetics of Voting: An Australian Twin Study." *Behavior Genetics* 37: 435–48.
- Heinz, A., et al. 2005. "Amygdala-Prefrontal Coupling Depends on a Genetic Variation of the Serotonin Transporter." *Nature Neuroscience* 8: 20–21.
- Hewitt, J. K. 2012. "Editorial Policy on Candidate Gene Association and Candidate Gene-by-environment Interaction Studies of Complex Traits." *Behavior Genetics* 42: 1–2.
- Hu, X.-Z., et al. 2006. "Serotonin Transporter Promoter Gain-of-Function Genotypes Are Linked to Obsessive-Compulsive Disorder." *American Journal of Human Genetics* 78: 815–26.
- Ioannidis, J. 2005. "Why Most Published Research Findings Are False." *PLoS Medicine* 2 (8): e124. doi:10.1371/journal.pmed.0020124.
- Jackson, R. A., and T. M. Carsey. 2007. "U.S. Senate Campaigns, Negative Advertising, and Voter Mobilization in the 1998 Midterm Election." *Electoral Studies* 26 (1): 180–95.
- Jankowski, Richard. 2007. "Altruism and the Decision to Vote: Explaining and Testing High Voter Turnout." *Rationality and Society* 19 (1): 4–34.
- Kan, K. H., and C. C. Yang. 2001. "On Expressive Voting: Evidence from the 1988 US Presidential Election." *Public Choice* 108 (3–4): 295–312.
- Karp, J. A., and D. Brockington. 2005. "Social Desirability and Response Validity: A Comparative Analysis of Overreporting Voter Turnout in Five Countries." *Journal of Politics* 67 (3): 825–40.
- King, G. 1998. *Unifying Political Methodology: The Likelihood Theory of Statistical Inference*. Ann Arbor: University of Michigan Press.
- Klemmensen, R., P. K. Hatemi, S. B. Hobolt, A. Skytthe, and A. S. Nørgaard. 2011. "Heritability in Political Interest and Efficacy across Cultures: Denmark and the United States." *Twin Research and Human Genetics* 15 (1): 15–20.
- Knack, Stephen. 1991. "Social Altruism and Voter Turnout: Evidence from the 1991 NES Pilot Study." 1991 NES Pilot Study Reports.
- Koch, J. 1998. "The Perot Candidacy and Attitudes toward Government and Politics." *Political Research Quarterly* 51 (1): 141–53.
- Lande, R. 1980. "The Genetic Covariance between Characters Maintained by Pleiotropic Mutations." *Genetics* 94: 203–15.
- Littvay, L. 2012. "Do Heritability Estimates of Political Phenotypes Suffer from an Equal Environment Assumption Violation? Evidence from an Empirical Study." *Twin Research and Human Genetics* 15 (1): 6–14.
- Loewen, P. J., and C. T. Dawes. 2012. "Heritability of Duty and Voter Turnout." *Political Psychology* 33 (3): 363–73.
- Lynch, M. 2010. "Rate, Molecular Spectrum, and Consequences of Human Mutation." *PNAS* 107 (3): 961–68.
- Mackay, T. F. 2001. "The Genetic Architecture of Quantitative Traits." *Annual Review of Genetics* 35: 303–39.
- Martin, N. G., et al. 1986. "Transmission of Social Attitudes." *PNAS* 83 (12): 4364–68.
- McDermott, R., et al. 2009. "Monoamine Oxidase A Gene (MAOA) Predicts Behavioral Aggression Following Provocation." *PNAS* 106 (7): 2118–23.
- McDermott, R. E., and P. K. Hatemi, ed. 2011. *Man Is by Nature a Political Animal: Evolution, Biology, and Politics*. Chicago: University of Chicago Press.
- Medland, S. E., and P. K. Hatemi. 2008. "Political Science, Biometric Theory and Twin Studies: A Methodological Introduction." *Political Analysis* 17 (2): 191–214.
- Meyer-Lindenberg, A., et al. 2006. "Neural Mechanisms of Genetic Risk for Impulsivity and Violence in Humans." *PNAS* 103 (16): 6269–74.
- Mondak, J. J., M. V. Hibbing, D. Canache, M. A. Seligson, and M. R. Anderson. 2010. "Personality and Civic Engagement: An Integrative Framework for the Study of Trait Effects on Political Behavior." *American Political Science Review* 104 (1): 1–26.
- Murphy, D. L., and K. P. Lesch. 2008. "Targetting the Murine Serotonin Transporter: Insights into Human Neurobiology." *Nature Reviews Neuroscience* 9: 85–96.
- Norton, E., and E. Han. 2008. "Genetic Information, Obesity, and Labor Market Outcomes." *Health Economics* 17 (9): 1089–1104.
- Olsen, M. E. 1972. "Social Participation and Voting Turnout: A Multivariate Analysis." *American Sociological Review* 37 (3): 317–33.
- Oxley, D. R., et al. 2008. "Political Attitudes Vary with Physiological Traits." *Science* 321 (5896): 1667–70.
- Pearson, T., and T. Manolio. 2008. "How to Interpret a Genome-wide Association Study." *Journal of the American Medical Association* 299 (11): 1335–44.
- Plomin, Robert. 2008. *Behavioral Genetics*. 5th ed. New York: Worth Publishers.
- Price, A., et al. 2009. "The Impact of Divergence Time on the Nature of Population Structure: An Example from Iceland." *PLoS Genetics* 5 (6): 1–10.
- Rhee, S. H., and I. D. Waldman. 2002. "Genetic and Environmental Influence on Antisocial Behavior: A Meta-analysis of Twin and Adoption Studies." *Psychological Bulletin* 128: 90–529.
- Risch, N., et al. 2009. "Interaction between the Serotonin Transporter Gene (5-HTTLPR), Stressful Life Events, and Risk of Depression:

- A Meta-analysis." *Journal of the American Medical Association* 301: 2462–71.
- Sallach, D. L., N. Babchuk, and A. Booth. 1972. "Social Involvement and Political Activity: Another View." *Social Science Quarterly* 52 (March): 79–92.
- Schwanhäusser, B. R. J., et al. 2011. "Global Quantification of Mammalian Gene Expression Control." *Nature* 473 (7347): 337–42.
- Settle, J. E., C. T. Dawes, N. A. Christakis, and J. H. Fowler. 2010. "Friendships Moderate an Association between a Dopamine Gene Variant and Political Ideology." *Journal of Politics* 72 (4): 1189–98.
- Settle, J. E., C. T. Dawes, and J. H. Fowler. 2009. "The Heritability of Partisan Attachment." *Political Research Quarterly* 62 (3): 601–13.
- Smith, K. B., J. R. Alford, P. K. Hatemi, L. J. Eaves, C. Funk, and J. R. Hibbing. 2012. "Biology, Ideology and Epistemology: How Do We Know Political Attitudes Are Inherited and Why Should We Care?" *American Journal of Political Science* 56 (1): 17–33.
- Smith, K. B., and P. K. Hatemi. 2012. "OLS is AOK for ACE: A Regression-based Approach to Synthesizing Political Science and Behavioral Genetics Models." *Political Behavior*. doi:10.1007/s11109-012-9192-2.
- Smolen, A., and J. K. Hewitt. 2003. "Biomarkers in Wave III of the Add Health Study." Technical Report, National Longitudinal Study of Adolescent Health.
- Smolen, A., et al. 2012. "Add Health Wave IV Documentation Report: Candidate Genes" Technical Report, National Longitudinal Study of Adolescent Health.
- Speliotes, E. K., et al. 2010. "Association Analyses of 249,796 Individuals Reveal 18 New Loci Associated with Body Mass Index." *Nature Genetics* 42: 937–48.
- Stam, A. C., A. Von Hagen-Jamar and A. B. H. Worthington. 2012. "Fear and Attitudes towards Torture and Preventive War." *Twin Research and Human Genetics* 15 (1): 60–70.
- Timpane, R. J. 1998. "Ties that Bind - Measurement, Demographics, and Social Connectedness." *Political Behavior* 20 (1): 53–77.
- Verhulst, B. 2012. "Integrating Classical and Contemporary Explanations of Political Participation." *Twin Research and Human Genetics* 15 (1): 42–51.
- Verhulst, B., & R. Estabrook. 2012. "Using Genetic Information to Test Causal Relationships in Cross-Sectional Data." *Journal of Theoretical Politics* 24 (3): 328–44.
- Verhulst, B., P. K. Hatemi, and L. J. Eaves. 2012a. "Covariation Not Causation: The Relationship between Personality Traits and Political Ideologies." *American Journal of Political Science* 56 (1): 34–51.
- Verhulst, B., P. K. Hatemi, and L. J. Eaves. 2012b. "Disentangling the Importance of Psychological Predispositions and Social Constructions in the Organization of American Political Ideology." *Political Psychology* 33 (3): 375–93.
- Verhulst, B., P. K. Hatemi, and N. G. Martin. 2010. "The Nature of the Relationship between Personality Traits and Political Attitudes." *Personality and Individual Differences* 49: 306–16.
- Vinkhuyzen, A. A. E., et al. 2012. "Common SNPs Explain Some of the Variation in the Personality Dimensions of Neuroticism and Extraversion." *Translational Psychiatry* 2: e102. doi:10.1038/tp.2012.27.
- Visscher, P. M., et al. 2006. "Assumption-Free Estimation of Heritability from Genome-wide Identity-by-descent Sharing between Full Siblings." *PLoS Genetics* 2: e41.
- Visscher, P. M., M. A. Brown, M. I. McCarthy, and J. Yang. 2012. "Five Years of GWAS Discovery." *American Journal of Human Genetics* 90: 7–24.
- Weber, C., M. Johnson, and K. Arceneaux. 2011. "Genetics, Personality, and Group Identity." *Social Science Quarterly*, 92 (5): 1314–37.
- Wilson, E. O. (1998). *Consilience: The Unity of Knowledge*. New York: Knopf.
- Xu, H., and S. Shete. 2006. "Mixed-effects Logistic Approach for Association Following Linkage Scan for Complex Disorders." *Annals of Human Genetics* 71: 230–37.
- Yang, J., et al. 2010. "Common SNPs Explain a Large Proportion of the Heritability for Human Height." *Nature Genetics* 42 (7): 565–69.