ADDRESSING THE ENDOGENEITY OF ECONOMIC EVALUATIONS IN MODELS OF POLITICAL CHOICE

MARK PICKUP*
GEOFFREY EVANS

Abstract  Subjective measures used in models of political choice are typically open to criticism with respect to their endogeneity. Economic perceptions have been subject to particular criticism in this respect. We address these concerns through the application of estimation procedures introduced from econometrics and the introduction of a new measure of economic perceptions that acts as an instrumental variable. Together, these eliminate simultaneity bias and bias due to unobserved heterogeneity, reduce omitted-variable bias, and reduce noise/measurement error in economic perceptions. An analysis of three-wave panel surveys produces estimates of the effect of economic perceptions on party evaluation that are not biased upward by the presence of endogeneity and that help address the discrepancy between competing subjective models of economic effects on government approval.

Introduction
Subjective measures are routinely used in models of political choice but are vulnerable to concerns regarding their endogeneity. The task of estimating effects purged of endogeneity involves not only operationalizing constructs effectively, but also estimating the nature of their association with outcome variables using appropriate sources of data and estimation procedures. Measurement therefore involves more than concept-indicator linkage; it requires the sources of bias and error in a particular hypothesized relationship to be specified and controlled. Our aim in this paper is to introduce methods for estimating two types of effects—“between individuals” and “within

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individuals over time”—which can be employed with appropriate estimation procedures and data to remove bias and error deriving from endogeneity.

To illustrate this approach, we focus on a debate in political science over economic voting that uses evaluations of the economy to account for patterns of voting, specifically punishment of the incumbent when the economy is perceived to be underperforming. Although this theory has been influential in providing both a heuristic for voter decision-making and a performance barometer of electoral accountability, a dispute has arisen over the extent to which individuals’ evaluations of the state of the economy in the survey context are endogenous to their partisan and other characteristics. We argue that this dispute can be advanced by delineating different solutions to the endogeneity problems for the two types of effects: (1) the effect of an individual’s subjective evaluations of the economy differing from those of another individual (differences across/between individuals); and (2) the effect of an individual’s evaluations changing over time (differences within individuals).

We proceed by identifying four distinct sources of endogeneity. We then introduce instruments, estimation procedures, and types of data that can account for these different sources of bias. To estimate the effects of differences between individuals, while controlling for endogeneity problems, we introduce a new measure and implement a two-stage least squares instrumental variable (IV) analysis with cross-sectional data (solution 1). We address endogeneity when estimating the effect of changes within individuals by using a likelihood-based estimator of a dynamic panel data model developed by Lancaster (2002). This analytic procedure is combined with the new measure to produce endogeneity-purged estimates of effects (solution 2).

The Substantive Issue: Explaining Political Preferences with Economic Evaluations?

Economic voting models give center stage to voters’ subjective evaluations of the economy as predictors of incumbent approval and vote choice. Economic conditions influence electoral outcomes as voters punish or reward incumbents for economic performance. Aggregate appraisals of the macro-economy have generally been assumed to drive presidential election outcomes (Erikson 1989; Erikson, MacKuen, and Stimson 2002), with retrospective perceptions of the national economy (typically, “How do you think the general economic situation in this country has changed over the past twelve months?”) motivating the individual voting decisions underlying them (Fiorina 1981; Lewis-Beck and Stegmaier 2007).

Recently, however, studies have argued that economic evaluations are powerfully influenced by endogenous partisan considerations and, as such, their effects on government approval and vote are likely to have been overestimated. Examples of this body of work include panel analyses (Evans and
Andersen 2006; Evans and Pickup 2010), which find that retrospective macro-
economic perceptions are strongly conditioned by one-year lagged opinions of
the incumbent party and have little or no independent effects. Before-and-after
election designs (Anderson et al. 2004; Ladner and Wlezien 2007) demon-
strate that economic expectations and retrospective evaluations are condi-
tional on voters’ estimates of election outcomes. Nonrecursive cross-sectional
models (van der Eijk et al. 2007; Wlezien, Franklin, and Twiggs 1997) find
that pre-election vote intention predicts simultaneously measured perceptions
of economic performance. Survey experiments (Wilcox and Wlezien 1993;
Tilley and Hobolt 2011) find strong evidence that questions about the econ-
omy contain implicit partisan cues, in addition to a broader range of evidence
on the biases, partisan and otherwise, affecting economic assessments (Duch,
Palmer, and Anderson 2000; Bartels 2002) and the influence of the partisan
cues on attributions of responsibility to incumbents for positive versus nega-
tive economic outcomes (Rudolph 2003; Marsh and Tilley 2010). As a con-
sequence, the role of economic appraisals in providing a fulcrum of electoral
accountability has been thrown into doubt (Anderson 2007). Does this mean
that we should not use economic perceptions in models of government evalua-
tion and vote choice? The answer rests upon the degree to which the economic
evaluations in these models are endogenous.

Identifying Sources of Endogeneity Bias

The term endogeneity encompasses a number of different problems. We iden-
tify four sources of endogeneity pertinent to economic voting and approval
models, each of which can result in biases when estimating such models, if
not addressed. Let

\[ y_{it} = \text{Government Approval} \]
\[ x_{it} = \text{Economic Evaluation}. \]

The theory and past empirical examinations considered above suggest that the
true data-generating processes for these two variables are

\[ y_{it} = \alpha_1 y_{i,t-1} + \beta_1 x_{i,t} + \sum_{m=2}^{M} \beta_{m1} z_{mi,t} + \mu_{1it} \quad (1) \]
\[ x_{it} = \alpha_2 x_{i,t-1} + \beta_{12} y_{i,t-1} + \sum_{l=2}^{L} \beta_{12} z_{li,t} + \mu_{2it} \quad (2) \]
\[ \mu_{jit} = \tau_i + \eta_{ji} + \epsilon_{jit} \quad (3) \]
\[ \epsilon_{jit} \sim NID(0, \sigma_j^2) \]
where $\tau_t$ represents a common trend across all cases, $\eta_{ji}$ represents time-invariant unobserved individual heterogeneity, and $\varepsilon_{jit}$ are independent, identically, and normally distributed (3). The data-generating process suggests that $x_{it}$ is a function of $y_{it-k}$ (2). The $z_{mit}$ are exogenous predictors of $y_{it}$, and the $z_{lit}$ exogenous predictors of $x_{it}$. Generally, $k$ in (2) is assumed to be either 0 or 1. This means that economic evaluations are a function of current government approval and/or a lag of government approval. The typical economic accountability model uses ordinary least squares regression to estimate

$$y_{it} = \beta_{01} + \beta_{11}x_{it} + \nu_{iit}$$

for $T = 1$. Other exogenous predictors are also usually included, in which case

$$\nu_{iit} = \mu_{iit} + \alpha_iy_{iit-k}$$

$$= \eta_{ii} + \varepsilon_{iit} + \alpha_iy_{iit-k}$$

As $T = 1$, this is a cross-sectional model and the exogeneity assumption for $x_{it}$ for unbiasedness in the estimation of $\beta_{11}$ is $E(\nu_{iit} | x_{it}) = 0$, and the exogeneity assumption for asymptotic unbiasedness as $N \to \infty$ is $\text{Cov}(\nu_{iit}, x_{it}) = 0$.\(^2\)

If (1) and (2) do represent the data-generating process, the OLS estimation of the parameters in model (4) contains the following endogeneity problems. (A formal explanation of how each source of endogeneity violates these assumptions is included in the online appendix, section A.1.)

The first source of endogeneity is a consequence of not including the lag of government approval ($\alpha_iy_{iit-1}$), which predicts current values of both economic evaluations (when $k = 1$) and government approval in the model. The model is not dynamically complete (Wooldridge 2006, 400–402). The second source of endogeneity is a consequence of not controlling for the fixed characteristics of individuals that predict both economic evaluations and approval. This is commonly known as an unobserved heterogeneity problem. The third source of endogeneity is a consequence of not controlling for time-varying characteristics of individuals that predict both economic evaluations and approval. This is commonly known as omitted-variable bias. The fourth source of endogeneity is a consequence of the contemporaneous direction of causality not running completely from economic evaluations to approval but also (maybe

1. The inclusion of $\beta_{01}$ in (4) means that the $\eta_{ii}$ will be mean centered. Further, for $T = 1$, $\tau_{it}$ is a constant and incorporated into $\beta_{01}$.
2. We acknowledge that specifying the conditions for exogeneity in this way, while common, can lead to an ambiguity when generalized (Hendry 2003, chapter 5; Engle, Hendry, and Richard 1983). However, we are specifying these conditions in the context of a specific model and a specific data-generating process. Therefore, we believe they suffice for our purposes.
exclusively) in the reverse direction (when \( k = 0 \)). This causes simultaneity bias. These sources of endogeneity are summarized in table 1.

**Potential Solutions to Endogeneity**

The economic voting model implies both within-individual over-time and between-individuals static relationships between retrospective economic evaluations and government approval. It is possible, but certainly not necessary, that the two relationships are the same. It is quite common when considering psychological processes to allow these to be different (Curran and Bauer 2011). Distinguishing between/across-individuals and within-individual variance and the differing effects of regressors on each is the rationale behind the hierarchical modeling of longitudinal data (Raudenbush 2001). It is the basis of much econometric work (Arellano 2003) and has a long pedigree (Lillard and Willis 1978).

To make the difference clear in the data-generating process, we can rewrite (1) as follows:

\[
y_{it} = \alpha_{1} y_{i, t-1} + \beta_{1,1,1} \overline{x}_i + \beta_{1,1,2} \left( x_{it} - \overline{x}_i \right) + \mu_{1,t}
\]  

(6)

The coefficient \( \beta_{1,1,1} \) captures the effects of average differences between individuals, and \( \beta_{1,1,2} \) captures the effects of differences within individuals over time. If (1) is estimated without making the distinction, the resulting estimate of \( \beta_{1,1,1} \) will be an average of the two effects (Greene 2003, 290). A common interpretation for the distinction is that the first is the effect of permanent (or at least long-term) differences in \( x_{it} \), and the second is the effect of short-term temporal fluctuations in \( x_{it} \). In our case, there may be long-term differences in economic evaluations across individuals. These differences may be due to differences in socialization or due to individuals from different classes and/or occupations experiencing different parts of the economy (e.g., Rehm 2009). There may also be short-term fluctuations in economic perceptions within individuals. These fluctuations need not be common across all individuals. They may be due to individuals reading news stories about the economy, viewing a campaign ad that makes reference to the economy, or discussing the economy with peers (e.g., Hetherington 1996; Sanders and Gavin 2004).

Regardless of which of these effects we want to test, we must take into account each of the four potential endogeneity problems. However, the solutions will differ: solution 1 will apply if we wish to test for the effect of differences between individuals, and solution 2 will apply if we wish to test for the effects of changes within individuals.
<table>
<thead>
<tr>
<th>Source</th>
<th>Example</th>
</tr>
</thead>
<tbody>
<tr>
<td>(Y_{it-1})</td>
<td>Both current government approval (Y_{it}) and current economic evaluations (X_{it}) are a function of past government approval (Y_{it-1}), resulting in a biased estimate of the causal relationship between current economic evaluations and current government approval.</td>
</tr>
<tr>
<td>(Z_{i, fixed})</td>
<td>Both current government approval (Y_{it}) and current economic evaluations (X_{it}) are a function of fixed characteristics of individuals (Z_{i, fixed}) (e.g., race, class), resulting in a biased estimate of the causal relationship between current economic evaluations and current government approval.</td>
</tr>
<tr>
<td>(Z_{it})</td>
<td>Both current government approval (Y_{it}) and current economic evaluations (X_{it}) are a function of current partisan identification (Z_{it}), resulting in a biased estimate of the causal relationship between current economic evaluations and current government approval.</td>
</tr>
<tr>
<td>(X_{it})</td>
<td>The direction of influence between current government approval (Y_{it}) and current economic evaluations (X_{it}) runs in both directions or from government approval to economic evaluations, resulting in a biased estimate of the causal relationship between current economic evaluations and current government approval.</td>
</tr>
</tbody>
</table>
We can test for the effect of differences between individuals with cross-sectional data and model (4). In these circumstances, a potential solution to the four endogeneity problems is to instrument $x_{it}$ (Wooldridge 2006). As is the case with any instrument, there are two specific requirements for the instrumental variable: (1) it must be partially correlated with $x_{it}$, controlling for the other exogenous variables in the model; and (2) it must not be correlated with the errors in the model, which include the problematic elements $y_{it-1}$ and $\eta_i$ (Angrist and Krueger 2001, 72–73; Wooldridge 2002, 89–90; Greene 2003, 76). This second requirement is equivalent to saying the instrumental variable is not correlated with $y_{it}$ after the exogenous variables in the model (instrumented or otherwise) are partialed out. In this context, this instrumentation addresses the four sources of endogeneity by only using variance in $x_{it}$ that is independent of the errors in the estimation of (4). The challenge of this approach is to define a measure that is independent of the errors in the model by being independent of $y_{it-1}$, $\eta_i$, and any excluded time-varying covariates, after we control for the exogenous time-varying covariates that are included in the model.

For this purpose, we introduce a measure of economic evaluations that inquires about respondents’ beliefs regarding the appropriateness of purchasing major goods given current economic conditions. This measure is part of a suite of indicators that we have tested for their properties as measures of individuals’ economic perceptions. It is derived from that used as the basis of a “climate for major purchase” index, which forms part of the consumer confidence measure reported monthly by GfK NOP. A question like the major-purchase variable is on the consumer confidence index of most countries. For example, in Canada, the Conference Board of Canada asks as part of its Index of Consumer Confidence: “Do you think that right now is a good or bad time for the average person to make a major outlay for items such as a home, car, or other major item?”; and in the United States, the University of Michigan Index of Consumer Sentiment (ICS) includes “About the big things people buy for their homes—such as furniture, a refrigerator, stove, television, and things like that. Generally speaking, do you think now is a good or bad time for people to buy major household items?” The website for the University of Michigan ICS also includes figures demonstrating that the responses to major-purchase questions track true consumer behavior very well.

Economists have been using responses to such questions to predict consumer behavior at the aggregate level for decades. The close relationship

5. Such questions are rarely asked in surveys that also ask for vote intention and/or government approval. Given the value of this question as an instrument, we encourage its inclusion in political surveys.
between responses to the consumer confidence questions and actual consumer behavior suggests that these questions are tapping into meaningful economic perceptions that do not derive from partisan cuing in the way more typically used sociotropic evaluations have been argued to do. Our focus on the major-purchase indicator derives from both considerations of its face validity—it inquires about decisions of genuine interest to most people without having any obvious implications for attributions of a political sort—and tests of its properties vis-à-vis those of other measures that have been adapted from the aggregate-level literature for use as individual-level instruments.

Work conducted by the Survey Research Center at the University of Michigan has demonstrated that responses to questions about major purchases, such as houses and cars, are strong predictors of actual purchase behavior (www.sca.isr.umich.edu). Given the strong connection between individual responses to the major-purchase question and actual purchase behavior, it would seem that there is a sizeable cost to having a biased view of whether it is a good time to make a major purchase. This is in contrast to general economic evaluations, which do not have such a cost. It is the real cost of having a biased view of whether it is a good time to make a major purchase (in contrast to general economic perceptions) that Gerber and Huber (2009) use to argue for the use of retail sales as a proxy to examine the relationship between economic perceptions and partisanship. The major-purchase question is intended to tap this form of exogenous economic evaluations.

The major-purchase indicator thus has prima facie plausibility as a potentially exogenous influence on political choices, and we demonstrate in the following section that the major-purchase measure, in an economic accountability model, passes statistical tests of exogeneity. We conclude that the major-purchase item does not exhibit the partisan influence of standard economic evaluations, solving the endogeneity problems of the traditional measures of sociotropic economic perceptions.

SOLUTION 2

Testing for the effect of changes within individuals requires us to use panel data and to estimate equation (1), accounting for the structure of the errors suggested by equation (3). All cross-case variance in government approval is
eliminated from the estimation of this model. Estimation is based on within-case (individual) variance and variation in that variance across individuals—e.g., some individuals’ approval increases, while that of other individuals decreases or remains the same.

The estimation presents a challenge. The OLS fixed-effects estimator with a lagged dependent variable is biased with a fixed (small) $T$. However, the Arellano and Bond GMM estimator (1991) and the Lancaster likelihood-based estimator (2002) are both viable. The application of the Arellano and Bond GMM estimator to this problem is discussed elsewhere (Evans and Pickup 2010). The likelihood-based estimator proceeds as follows. The maximum likelihood estimation of either model (1) or (2) with a fixed (small) $T$ leads to an incidental parameters problem (Neyman and Scott 1948). With fixed $T$, the number of fixed effects ($\eta_{ji}$) approaches infinity at the same rate as $N \to \infty$. Therefore, we cannot rely on asymptotics as $N \to \infty$, and the application of maximum likelihood leads to inconsistent estimators. Lancaster (2002) suggested a solution to this particular incidental-parameter problem.

The key to this approach is that we are not actually interested in estimates of the incidental parameters. We are interested in estimates of the “common parameters”—the effects of the dynamic valence considerations, such as economic evaluations. Therefore, we seek a re-parameterization of the incidental parameters in the likelihood so that the incidental and common parameters are information-orthogonal. This puts us in a position to produce a likelihood-based estimate that is independent of the values of the incidental parameters (i.e., we do not estimate the incidental parameters). This estimate is consistent as $N \to \infty$. Therefore, we continue to have incidental parameters but not an incidental-parameter problem. Further details are provided in the online appendix, section A.2.

The inclusion of $\alpha_{y_{t-1}}$ in the (1) removes it from the error term, resolving endogeneity problem 1. Re-parameterization of the fixed effects, $\eta_{ji}$, so that they are information-orthogonal in the likelihood, allows us to estimate the parameters of interest independent of them. This controls for the unobserved heterogeneity produced by these fixed effects and resolves endogeneity problem 2. This distinguishes solution 2 from solution 1, in which the problems of moving the lag of $y_{it}$ and the fixed effects into the error are resolved with an exogenous instrument. For solution 2, endogeneity problems 3 (omitted-variable bias) and 4 (simultaneity bias) remain. Omitted-variable bias can be reduced by including additional dynamic covariates, but there is never a guarantee that all relevant covariates are included. Alternatively, we can combine this panel data approach with the instrumental variable approach of solution 1, to ensure that we have addressed all four sources of endogeneity, while estimating the within-individual effect.

10. The code for implementing the likelihood-based approach in WinBUGS is available at http://pollob.politics.ox.ac.uk/.
Data and Measurement

Data designed to test these ideas were collected prior to and during the 2010 UK general election. We contracted the data collection as part of the British Cooperative Campaign Analysis Project. The BCCAP is a multiple-wave Internet panel study. Data were collected in September 2009, January 2010, and April 2010—approximately four months apart. Data collection was undertaken by the polling company YouGov (see the online appendix, section A.3, for full details).

YouGov uses targeted quota sampling from a respondent panel database to conduct web-based surveys. Its proprietary software looks at all surveys that currently need panel members, and calculates how many people to send invitations to every 30 minutes. Due to the way jobs are sampled, there is no survey response rate; however, the overall cooperation rate for the panel is 21 percent, with the average response time for a clicked e-mail being 19 hours from the point of sending. YouGov uses a rim weighting system (Deming 1985), which employs an iterative process to ensure that the sample has the correct proportions for each of the major demographics (or rims). To obtain a sample that reflects national population characteristics, the data are weighted to the profile of all adults aged 18+, taking into account age, gender, social class, region, party identification, and newspaper readership. Target percentages are derived from census data and the National Readership Survey.11

MEASURES

The BCCAP panel contains the standard retrospective assessment of the national economic situation, or “sociotropic” evaluation (Kinder and Kiewiet 1981), that has been the most commonly employed indicator used to estimate economic effects in the political science literature. More specifically, “retrospective sociotropic” models of economic voting posit that voters care about generalized social utility rather than their personal utility, and assess the provision of such utility on the basis of recent economic performance. Retrospective sociotropic economic evaluations have been found to have strong effects on government approval and vote (for reviews, see Anderson [2007]; Lewis-Beck and Stegmaier [2007]).

11. The National Readership Survey is a random probability survey comprising 34,000 random face-to-face interviews conducted annually. It is a continuous survey, conducted twelve months of the year, seven days a week. It has a sample of 36,000 face-to-face interviews a year with British adults aged 15+. The sample is a random probability sample with interviews only conducted at randomly selected addresses with randomly selected individuals. Those interviews are conducted in the respondent’s own home, and the average interview takes 27 minutes. Respondents are asked about their readership of a list of newspapers, newspaper supplements, and magazines, as well as a host of socio-demographic information about themselves. For more information, see http://www.nrs.co.uk.
As noted earlier, the retrospective sociotropic question is worded “How do you think the general economic situation in this country has changed over the past twelve months?” Response options are “got a lot worse”; “got a little worse”; “stayed the same”; “got a little better”; “got a lot better.” Henceforth, this measure is referred to as *retrospective economic evaluations*.

The BCCAP panel also contains the measure asking respondents whether it is a good time to make a major purchase that we use to instrument economic perceptions.\(^{12}\) Henceforth, we refer to this question as *major purchase*. The BCCAP panel also contains a typical approval question for the governing party, asking respondents to indicate how they feel about the Labour Party on a 0 (unfavorable) to 10 (favorable) scale. This forms the dependent variable for the analysis of party choice.

Finally, we include several standard control variables that are typically included in economic accountability models and which are available in the BCCAP surveys:\(^{13}\) These are middle-class (1), other (0); union member (1); homeowner (1); male (1); university graduate (1). Middle-class is defined as categories A, B, and C1 of the National Readership Survey Social Grade classification. This classification is based on occupation (http://www.nrs.co.uk/lifestyle-data/). Response categories for income are under £5000 per year; £5000 to £9999 per year; £10,000 to £14,999 per year; £15,000 to £19,999 per year; £20,000 to £24,999 per year; £25,000 to £29,999 per year; £30,000 to £34,999 per year; £35,000 to £39,999 per year; £40,000 to £44,999 per year; £45,000 to £49,999. Partisan identification is classified as (1) Labour, (0) other party or none.

**Results**

For solution 1, we use the responses to the major-purchase question as an instrument for the usual retrospective economic evaluation variable. We begin by running a two-stage least squares instrumental-variable regression for model (4). We include the demographic control variables listed above.

The analysis is run on the last wave of data (April). A common approach to between-individual estimation is to run a cross-sectional analysis using the cross-wave average for each variable. This reduces the noise from random, individual-level deviations from their long-term values. However, the last wave is special in that it is the election wave. In most instances, researchers have data only from this wave, and this is the point in time that they are typically most interested in examining.

Both the current and change (since the January wave) values of the major-purchase variable are used to instrument the retrospective economic evaluation variable. The intuition behind the use of the change in the major-purchase variable as an instrument is that the retrospective economic evaluation question that

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12. The wording for the survey questions is provided in the online appendix.
13. Descriptive statistics for each of these variables are provided in the online appendix, section A.4.
it instruments asks for an assessment of how the economy has changed. The use of two instruments allows us to apply diagnostic tests that require multiple instruments, but the results from instrumenting only with current values are substantively the same (see the online appendix, section A.5). Heteroskedasticity-robust standard errors are calculated. The results are presented in table 2.

The results suggest that the instrumented economic evaluations are a predictor of approval. Before interpreting this further, we test whether the major-purchase variables are weak instruments for retrospective economic evaluations. This is done by testing the null hypothesis that the major-purchase variables have no significant explanatory power for retrospective economic evaluations after controlling for the other variables included in the model. We do this using an $F$-statistic: $F(2,509) = 10.06 \ (p = 0.0001)$. We can reject the null hypothesis that the current and change major-purchase variables are weak instruments.

Table 2. Instrumental Variables (SLS) Estimates of Cross-Sectional Model (robust standard errors in parentheses)

<table>
<thead>
<tr>
<th></th>
<th>Approval with instrumented economic evaluations</th>
<th>Approval with retrospective economic evaluations</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$N = 520$</td>
<td>$N = 520$</td>
</tr>
<tr>
<td>Economic evaluations</td>
<td>1.37**</td>
<td>0.93**</td>
</tr>
<tr>
<td></td>
<td>(0.51)</td>
<td>(0.10)</td>
</tr>
<tr>
<td>Middle-class</td>
<td>0.29</td>
<td>0.24</td>
</tr>
<tr>
<td></td>
<td>(0.23)</td>
<td>(0.23)</td>
</tr>
<tr>
<td>Union</td>
<td>0.90**</td>
<td>0.86**</td>
</tr>
<tr>
<td></td>
<td>(0.26)</td>
<td>(0.24)</td>
</tr>
<tr>
<td>Homeowner</td>
<td>-0.088</td>
<td>-0.17</td>
</tr>
<tr>
<td></td>
<td>(0.30)</td>
<td>(0.28)</td>
</tr>
<tr>
<td>Male</td>
<td>0.089</td>
<td>0.12</td>
</tr>
<tr>
<td></td>
<td>(0.20)</td>
<td>(0.19)</td>
</tr>
<tr>
<td>Age (years)</td>
<td>-0.0073</td>
<td>-0.010</td>
</tr>
<tr>
<td></td>
<td>(0.0085)</td>
<td>(0.0075)</td>
</tr>
<tr>
<td>University</td>
<td>-0.35</td>
<td>-0.18</td>
</tr>
<tr>
<td></td>
<td>(0.30)</td>
<td>(0.22)</td>
</tr>
<tr>
<td>Income category</td>
<td>-0.064</td>
<td>-0.051</td>
</tr>
<tr>
<td></td>
<td>(0.034)</td>
<td>(0.031)</td>
</tr>
<tr>
<td>PID (Labour)</td>
<td>3.25**</td>
<td>3.59**</td>
</tr>
<tr>
<td></td>
<td>(0.50)</td>
<td>(0.25)</td>
</tr>
<tr>
<td>Intercept</td>
<td>8.30**</td>
<td>6.78**</td>
</tr>
<tr>
<td></td>
<td>(1.83)</td>
<td>(0.57)</td>
</tr>
<tr>
<td>$R^2$</td>
<td>0.49</td>
<td>0.51</td>
</tr>
</tbody>
</table>

*aEconomic evaluations instrumented with current and change values of major purchase and included exogenous variables.

**$p < .01$
We can now test whether or not the instruments are correlated with the error term. Specifically, we use Wooldridge’s heteroskedasticity-robust score test of overidentifying restrictions. The null hypothesis is that the instruments are uncorrelated with the error term. This is the second of the IV assumptions and is known as an exclusion restriction (Wooldridge 2006, 525). The score test has a chi-square distribution (\(\chi^2(1) = 0.76; p = 0.38\)). The score indicates that we cannot reject the null hypothesis that the instruments are uncorrelated with the error term.14

The difficulty with this or any test of the exclusion restriction is that when the instruments pass the overidentification test, it could very well mean the instruments are uncorrelated with the error term (they are exogenous), but there does remain the possibility that the instruments pass this test because both instruments are endogenous in a way that produces biases of the same magnitude (Wooldridge 2002). This is a real difficulty with cross-sectional data and is one of the elements in the compelling argument made by some that an instrument is rarely convincing outside experimental randomization (e.g., Savoy and Green 2008). However, much can also be done with longitudinal data to support our overidentification test.15

If one instrument is convincingly exogenous, the overidentification test will not have the above weakness. Our two instruments are major purchase and the change in major purchase. The potential sources of endogeneity for our instruments are equivalent to those for the original economic evaluation variable: (1) the instrument is predicted by a lag of government approval; (2) the instrument is predicted by fixed characteristics of individuals that also predict government approval; (3) the instrument is predicted by omitted time-varying variables; or (4) the instrument is caused by government approval (reverse causation). As we demonstrate when we present the solution 2 results below, we can use the panel data to demonstrate that changes in major purchase are independent of changes in government approval. Thus, the change instrument is unlikely to be endogenous to government approval due to any time-varying omitted variables (endogeneity problem 3) or reverse causation (problem 4). Since the instrument represents a change in the individuals’ opinions, it is also unlikely to be endogenous due to any fixed characteristics (problem 2). These elements are differenced out in the major-purchase change variable. Further, we use the panel data to demonstrate that change in major purchase is not predicted by a lag of government approval and so will not be endogenous due to the exclusion of the lag of government approval in model 4 (problem 1). These facts provide substantial support for our contention that the major-purchase change instrument is exogenous and that the results of the overidentification test indicate that the major-purchase variable is also exogenous.

14. Further tests of the major-purchase variable as an instrument are included in the online appendix, section A.6.
15. Savoy and Green (2008) do suggest that a lag of a variable might be a convincing instrument.
Having rejected the null hypothesis that the major-purchase variables are weak instruments and having not rejected the null hypothesis that the instruments are uncorrelated with the error term, we can proceed with interpreting the results from the IV estimation. Economic evaluations are estimated on a five-point scale, and government approval on a ten-point scale. The effect of a one-unit difference in economic evaluations is a 1.37-unit difference in government approval.

Table 2 also presents the results of estimating the same model without instrumenting retrospective economic evaluations. We would still have found their effect to be statistically significant but would have concluded that they are smaller in magnitude. The discrepancy is probably because our instrumental variables are effective at filtering measurement error noise out of the economic evaluations measure (Wooldridge 2002, 95). The noise may be creating attenuation bias in the OLS estimation of the model based on equation (4).

We now examine solution 2 with the application of the Lancaster likelihood-based approach to estimating the parameters in a panel model based on equation (1). This estimates the effects of changes in economic evaluations within individuals. To allow for the potential of a common trend across all cases, as indicated by $\tau$ in equation (3), we explicitly estimate these parameters. We originally included a nonlinear trend by allowing for a different intercept in each wave (including $\tau_2$ and $\tau_3$ with wave one as the control) but found that the results are indistinguishable from using a linear trend (including $\tau_2$ and $2\tau_2$ with wave one as the control). These estimates are presented in the first two columns of Table 3.

The results indicate that individual changes in economic evaluations did have a statistically significant and positive effect on government approval from 2009 until the 2010 general election. However, these results control only for endogeneity problems 1 and 2. To control for endogeneity problems 3 and 4, we instrument retrospective economic evaluations.

The results of instrumenting retrospective economic evaluations in our panel data model are presented in the third column of Table 3. Before interpreting the results, we examine the strength of the instrumental variable in the panel setting. From the model estimation, we get individual tests of the null hypothesis that the instrument does not predict retrospective economic valuations in each wave (see the online appendix, section A.6). In all three waves, we can reject the null hypothesis. The instruments correlate in a significant way with the instrumented variable.

As indicated in the presentation of the solution 1 results, we can also test the extent to which changes in our instrument are determined by government approval by estimating a panel model equivalent to equation (2)—assuming

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16. As the Lancaster approach is essentially Bayesian, it was straightforward to incorporate the imputation of missing values from all variables with the observed values from the same variables from other waves. For missing data values, imputed values are drawn from the posterior distribution of the variables conditioning on the observed values from the other periods.

17. The online appendix, section A.7, includes estimates of the model based on (2) using the Lancaster likelihood-based approach.

18. The online appendix, section A.2, provides detail on the estimation approach.
Table 3. Likelihood-Based Estimates of Panel Models†

<table>
<thead>
<tr>
<th></th>
<th>Approval with retrospective economic evaluations</th>
<th>Approval with retrospective economic evaluations</th>
<th>Approval with instrumented economic evaluations</th>
<th>Major purchase with approval</th>
<th>Major purchase with lag approval</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$T = 3N = 1,129$</td>
<td>$T = 3N = 1,129$</td>
<td>$T = 3N = 1,129$</td>
<td>$T = 3N = 1,129$</td>
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<tr>
<td>$\alpha_j$</td>
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<td>0.00001</td>
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<td></td>
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<td>(0.0019)</td>
<td>(0.002)</td>
<td>(0.0003)</td>
<td>(0.0003)</td>
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<tr>
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<td>0.09*</td>
<td>0.002</td>
<td>0.004</td>
<td>0.00009</td>
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<tr>
<td></td>
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<td>(0.03)</td>
<td>(0.06)</td>
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<td>(0.0003)</td>
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<td>1.04**</td>
<td>1.04**</td>
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</tr>
<tr>
<td></td>
<td>(0.02)</td>
<td>(0.02)</td>
<td>(0.017)</td>
<td>(0.01)</td>
<td>(0.01)</td>
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<tr>
<td>$\tau_2$</td>
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<td>0.21**</td>
<td>0.23**</td>
<td>-0.04**</td>
<td>-0.04**</td>
</tr>
<tr>
<td></td>
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<td>(0.03)</td>
<td>(0.028)</td>
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<tr>
<td>$\tau_3$</td>
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</table>

†These are Bayesian estimates derived using MCMC methods. The reported coefficients represent the medians of the posterior distributions. The standard deviations of the posterior distributions are in parentheses.

* $p < .05$; ** $p < .01$
first $k = 0$ and second $k = 1$—substituting the major-purchase variable for the retrospective economic evaluation variable. The results are presented in the last two columns of table 3.

The results indicate that within-individual changes in major-purchase evaluations are not determined by within-individual changes in government approval and that within-individual changes in major-purchase evaluations are not determined by past (lag of) major-purchase evaluations. Recall that this has important implications for solution 1. The first result provides strong support for the claim that the change instrument is not endogenous to government approval due to any time-varying omitted variables or reverse causation. The second result indicates that that the change instrument will not be endogenous due to any correlation with the lag of government approval.

Turning back to solution 2, these results demonstrate the importance of controlling for all forms of endogeneity and the value of our instrumental variable. The estimated effect of economic evaluations on government approval, once economic evaluations are instrumented, drops to statistical insignificance (table 3, column 3). Not only is the effect statistically insignificant, but the estimated coefficient is less than one-twentieth the magnitude of the non-instrumented estimate.

Discussion

We have identified sources of endogeneity and specified solutions that can appropriately address them. Our first solution introduced the use of an instrumental variable approach to reduce bias and noise/measurement error when estimating the effect of differences in economic evaluations between individuals in cross-sectional data. However, the effect of differences between individuals is only one potentially observable implication of economic voting theory. When an individual’s evaluation of the economy changes, their support for the incumbent party is also expected to change. Identifying this effect requires panel data, and thus the second solution introduced the likelihood-based estimation of a panel model developed by Lancaster (2002). The estimated panel model provides a useful estimate of endogeneity-purged economic effects that deal with source of endogeneity 1 and 2. Although endogeneity problem 3 can also be reduced using this method through the addition of relevant dynamic variables in model (1), it cannot resolve endogeneity problem 4. Therefore, we proposed that the likelihood-based estimation of the panel model be used in conjunction with the instrument for subjective economic evaluations, thereby also addressing endogeneity problems 3 and 4.

On the basis of this analysis, it appears that responses to the “major purchase” question serve to isolate those aspects of a respondent’s retrospective economic evaluations that are actually driven by economic performance from those that derive from factors such as partisan conditioning. Our instrument passes the usual test of being a strong instrument and being uncorrelated with
the model errors, whether modeling cross-sectional differences or over-time changes in government approval. We also go further than the usual cross-sectional tests by using the availability of panel data to demonstrate the validity of our instrument.

Our use of an Internet panel leaves open the generalizability of these findings to the general population, though it is unlikely to do so any more than any other panel, plagued as all panels are by attrition that weakens their representativeness with respect to, typically, less politically interested respondents. Moreover, as reported in the online appendix, section A.3, the validation exercises conducted on the YouGov procedures indicate equivalent voting model parameters to those obtained from a cross-sectional face-to-face random sample (Sanders et al. 2007).

What implications does our analysis have for understanding the impact of economic evaluations on government approval and, ultimately, voting? We have seen that the application of the instrumental variable method to cross-sectional data indicates that long-term differences in economic evaluations between individuals do have an effect that is not endogenous. The estimated effect is similar to that of others using subjective economic indicators (van der Eijk et al. 2007; Duch and Stevenson 2008). Using the instrumental variable in conjunction with Lancaster’s orthogonal re-parameterization likelihood-based estimation procedure with panel data indicates that short-term changes in the economic evaluations of individuals do not have an exogenous effect. This is consistent with other panel data estimates of within-individual effects (Evans and Andersen 2006; Evans and Pickup 2010). These results resolve the differences in the findings across these studies. Economic evaluations do matter, but short-term changes in those evaluations do not. If and when we have comparable panel data for a longer period of time, we may well find that changes in the long term do have an exogenous effect on government support. Until then, we can conclude that voters are not responding to the small ripples in their evaluations of the economy, but it remains to be seen whether or not they respond to the long-term tide of economic sentiment.

Appendix. Question Wording

In addition to the sociotropic, retrospective economic evaluation question mentioned in the text, the following questions were asked as part of the British Cooperative Campaign Analysis Project survey.

**Governing Party Approval**

On the scale below, please indicate how you feel about the Labour Party. 0 < Unfavorable; 10 > Favorable; Don’t know.

**Major Purchase**

In view of the general economic situation, do you think now is the right time for people to make major purchases, such as furniture or electrical goods?
Now is the right time for people to make a major purchase; Now is neither the right nor the wrong time for people to make a major purchase; Now is the wrong time for people to make a major purchase.

**Homeowner**
Which of the following best describes your homeownership status?
Own outright; Own with a mortgage; Rent from a private landlord; Rent from a local authority/housing association; Other; Don’t know.

**Partisan Identification**
Generally speaking, do you think of yourself as Labour, Conservative, Liberal Democrat, or what?
Labour; Conservative; Liberal Democrat; Scottish National Party (SNP); Plaid Cymru; Green Party; United Kingdom Independence Party (UKIP); British National Party (BNP); Other; None; Don’t know.

Education, union membership, gender, and class were recorded previously by YouGov.

**Supplementary Data**
Supplementary data are freely available online at [http://poq.oxfordjournals.org/](http://poq.oxfordjournals.org/).

**References**


