Using Mechanical Turk as a Subject Recruitment Tool for Experimental Research

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We investigate Amazon.com's Mechanical Turk (MTurk) interface as a means for performing experimental political science research. After describing the MTurk interface and its promise as vehicle for performing low cost and easily fielded experiments, we investigate sample characteristics. Benchmarking MTurk respondents to important published experimental work in political science, we show that respondents recruited in this manner are more representative than in-person convenience samples—the modal sample in published experimental political science but modestly less representative than subjects in internet-based panels or national probability samples. We investigate two important concerns about MTurk samples: whether they are unengaged or frequent subjects. Additionally, we provide a tool to conduct panel surveys using the MTurk interface, discuss panel retention rates, and demonstrate that MTurk subjects are vastly less expensive than anything other than uncompensated student samples. In recent years, there has been increased interest in experiments in political science (citations). But experiments are neither cheap nor easy to implement, particularly when those experiments involve non-student adult subjects. Amazon.com's Mechanical Turk (MTurk) has the potential to facilitate low-cost experiments in political science with a diverse subject pool.

MTurk is an online web-based platform for recruiting and paying subjects to perform tasks. The benefit of using MTurk for subject recruitment is clear, Relative to other experimental pools, MTurk is extremely cheap; it is possible to run studies where subjects are paid as little as 10 or 15 cents. Not surprisingly, scholars across the social sciences have taken an interest in using MTurk to recruit research subjects to participate in online surveys, particularly surveys with embedded experimental manipulations.¹

However, despite this burgeoning line of research, the benefits and potential limitations of subject recruitment and research using MTurk remain relatively unexplored (though see Buhrmester, Kwang, and Gosling In press; Horton, Rand, Zeckhauser 2010).² This paper answers a simple but important question: Is Amazon's *Mechanical Turk* a valid tool for conducting research on the nature of public opinion and related questions in political science?

Our answer is a qualified "yes." In particular, the demographic characteristics of MTurk users, we show, are more representative and diverse than the corresponding student and convenience samples typically used in experimental political science studies. Although MTurk samples do not fare as well against high-quality and expensive representative samples, the

¹ At the time of writing, Google Scholar lists 231 social sciences articles with the phrase "Mechanical Turk." Relevant studies by economists include, e.g., Chandler and Kapelner 2010, Chen and Horton 2010; Horton and Chilton 2010. Computer scientists have also tested MTurk's suitability as a source of data for training machine learning algorithms (e.g., Sheng et al., 2008; Sorokin and Forsyth, 2008). For example, Snow et al. (2008) assessed the quality of MTurkers responses to several classic human language problems, finding that the quality was no worse than the expert data that most researchers use.

 $^{^{2}}$ Analyses have generally found that experiments on Internet samples yield results similar to traditional samples. Based on a comprehensive analysis, for example, Gosling et al. (2004) conclude that Internet samples tend to be diverse, are not adversely affected by nonserious or repeat responders, and produce findings consistent with traditional methods.

MTurk samples are – in informal terms – at least as representative, if not more so, than the convenience samples typically used in political science experiments.

We begin by providing an overview of the subject-recruitment and research gathering choices involved in using MTurk. Next, we describe results from a series of surveys we undertook to measure MTurk subjects' demographic and political characteristics. We compare our MTurk sample to those used in experiments published in leading political science journals, to a high-quality, internet panel sample, and to random-stratified samples used in the Current Population Survey (CPS) and the American National Election Studies (ANES). We demonstrate that the effects of experimental manipulations observed in the MTurk population comport well with those found in other studies. We also address two concerns raised about on-line samples: Whether sample pools are dominated by subjects who participate in numerous experiments (or participate more than once in a given experiment) and whether those subjects are effectively engaged with the survey stimuli. Finally, we describe methods for using MTurk to implement a panel survey. We show that it is possible to recontact respondents through MTurk at reasonably high rates and at minimal costs.

Recruiting Experimental Subjects via Mechanical Turk

A core problem for experimental researchers in political science is the difficulty and prohibitive cost of recruiting adult subjects. The high cost of subject recruitment has several undesirable effects, including a tendency to rely on convenience samples that are often dominated by student populations, which raise important external validity concerns about whether observed relationships would persist for non-student samples (Sears 1986, though see Druckman and Kam 2010). These concerns have led to important innovations to enable broader access to representative populations for experimental work, including the NSF-funded Time-

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Sharing Experiments in the Social Sciences (TESS) initiative, but access to these resources remains limited.

Amazon.com's Mechanical Turk is a promising vehicle for experimental subject recruitment. Amazon.com markets MTurk as a means to recruit individuals to undertake tasks In practice, those tasks involve a wide array of jobs requiring human intelligence, such as coding the attributes of product pictures for Amazon.com or transcribing handwriting, but they can also include taking surveys with embedded experimental manipulations at extremely low cost – as little as 10 cents per subject for a 5-minute experiment.

To initiate a survey using MTurk, a researcher – or, to use Amazon's language, "Requester" – establishes an account (www.mturk.com), places funds into her account, and then posts using the MTurk web interface a "job listing" that describes the Human Interaction Task (HIT) to be completed and the compensation to be paid. At this stage, the researcher can set requirements for subjects, including country of residence and "approval rate" – the percent of prior HITs submitted by the respondent that were subsequently accepted by Requesters. When a MTurk "worker" who meets these eligibility requirements logs onto their account, they can review the list of HITs available to them and choose to undertake any task for which they are eligible.

The MTurk interface gives the researcher a great deal of flexibility with which to conduct a study. In addition to using MTurk's embedded workspace to set up simple tasks, the researcher can also refer subjects to an external website to perform tasks – for instance, taking a survey with an embedded experimental manipulation. If subjects are referred to an external website, the site can be programmed to provide the subject with a unique code to enter in the MTurk website to

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demonstrate that they have completed the task.³ Additionally, outside websites make it easy to condition participation on informed consent, to debrief after a manipulation, and to collect detailed information about the survey process, including response times for items and respondents' location when taking the survey (on the basis of the MTurker's Internet Protocol, or IP, address).

The final stage for the researcher is compensating subjects. If the subject has entered a valid code from an external website, the researcher can simply authorize payment for the task through the MTurk interface. Additionally, if the researcher has arranged for the external website to produce a unique identifier, she can also now review that particular worker's task behavior to assess whether it is of the necessary quality. For example, if the experiment included mandatory filter questions or questions designed to verify the subject was reading instructions, the worker's compensation can be made contingent on responses to those questions.

Benchmarking MTurk Samples

MTurk makes it possible to easily conduct experimental studies. At the same time, we presently know little about the types of subjects who are available through MTurk relative to other subject pools. To assess the value for research of the MTurk subject pool we compare measured characteristics of MTurk survey participants to those characteristics in three distinct types of research samples: (a) convenience samples used in experiments published in leading political science journals, (b) a sample generated by a high-quality internet panel, and (c) probability samples of U.S. residents.

³ We have successfully used websites like SurveyGizmo and Qualtrics for this process, and any web survey service that can produce a unique worker code should perform equally well.

To learn about the characteristics of the MTurk population, we interviewed 587 individuals in February and March of 2010. We advertised the survey as taking about 10 minutes and paid respondents 50 cents each. Because we are benchmarking MTurk against samples of adult U.S. citizens, we restricted the survey to individuals MTurk classified as 18 or older and living in the U.S.⁴ We also excluded individuals with approval rates below 95 percent on previous MTurk tasks. As an additional check on U.S. residency, we verified that respondents took the survey from U.S. IP addresses and excluded 32 individuals (5.8 percent) who did not.⁵

Comparison of Respondent Characteristics: Local Convenience Samples

Local convenience samples are the modal means of subject recruitment among recent published survey and lab experimental research in political science. In particular, we examined all issues of the *American Political Science Review*, the *American Journal of Political Science*, and the *Journal of Politics* from January 2005 to June 2010. Of the 961 articles in these issues, 51 used experimental data. Forty-four of these articles used U.S. subjects exclusively (the complete list of these articles, as well as a summary of subject recruitment methods, appears in Appendix 1). Of these 44 articles, more than half used convenience samples for subjects (including student samples, local intercept samples, or temporary agencies).

Table 1 compares our MTurk sample to several convenience samples. After displaying the demographics of our MTurk sample, it displays the average characteristics from student samples collected by Kam, Wilking, and Zechmeister (2007). Next, the table lists characteristics of two adult convenience samples used in Berinsky and Kinder (2006) – one of the handful of articles that describes the characteristics of its convenience samples. One of these samples is

⁴ MTurk classifies individuals as 18 or older based on self reports. MTurk does not reveal how it classifies individuals as living in a particular country, but may rely on mailing addresses and credit card billing addresses. ⁵ These individuals may reside in the U.S., but be traveling or studying abroad. Additionally, although IP address locators seem reliable, we are unaware of research benchmarking their accuracy. Still, so as to provide as conservative a picture of our sample as is possible, we excluded these questionable respondents.

from around Princeton, New Jersey and the other is from around Ann Arbor, Michigan. Finally, to compare all of these to representative samples, the table also includes statistics from the Current Population Study (CPS) and the 2008 American National Elections Study (ANES). These latter two studies use face-to-face probability samples and are widely considered the "gold standard" for general opinion poll sampling.

In terms of demographic representativeness, the MTurk sample fares well in comparison with these convenience samples, especially the student sample.⁶ Not surprisingly, relative to an average student sample, the MTurk population is substantially older, but it is younger than the two Berinsky and Kinder samples. Focusing on the Berinsky and Kinder samples, the skew in the gender distribution of those samples was no better than the MTurk sample and, in fact, was worse in the Ann Arbor sample. More importantly for the purposes of political science experiments, the party identification skew was almost identical in the MTurk and Ann Arbor samples. Of course, our point is not to single out Berinsky and Kinder – the distribution of relevant demographic and political variables in their study was, in fact, more representative than that found in several other studies. Moreover, many other studies do not report any information about sample characteristics. Instead, we simply wish to emphasize that, when compared to the practical and publishable alternatives, the MTurk respondent pool has attractive characteristics – even before considering issues of cost.

Comparison of Respondent Characteristics: Knowledge Networks/ANESP

Apart from local convenience samples, the other dominant form of sample selection in the published experimental studies discussed above is internet-based surveys. Since political scientists are increasingly using existing Internet panels from which to draw subjects, we

⁶ Other researchers have surveyed MTurk respondents and found a similar demographic profile (e.g., Ross et al. 2010).

benchmark MTurk against a high-quality Internet survey, the American National Election 2008-2009 Panel Study (ANESP). The firm Knowledge Networks conducted the ANESP by recruiting respondents through a random-digit-dial method for a 21-wave internet-based panel survey (10 waves of the survey concerned political matters; the other 11 waves did not).⁷ Since we are treating the ANESP not as a best estimate of true population parameters, but rather as an example of a high-quality Internet sample, we present unweighted results from this survey. Comparing our MTurk survey with the ANESP has an additional advantage. Since both are Internet surveys, we hold the "mode" of survey constant. Additionally, in designing the MTurk survey, we followed the ANESP as closely as possible, using identical wordings and branching formats.

On many demographics, the MTurk sample is not substantially less representative of the population than the ANESP. These detailed demographic comparisons appear in Table 2. MTurk is only slightly more female, 60 versus 58 percent, and only slightly more educated, 14.9 versus 14.5 years. On race, MTurk's characteristics are mixed: It is closer to the CPS on percent white than is the ANESP, but is considerably worse on percent black, four versus nine percent. MTurk fares worse on age-related demographics, such as marital status, homeownership, and religious preference.⁸

On key political measures, MTurk subjects fair reasonably well by comparison with the ANESP. We present analysis for different outcomes in Table 3-Table 5. They register at about the same rate as do those in the 2008 ANES and CPS, about 80 percent, while respondents in the ANESP report a much higher 92 percent registration rate (Table 3). Among registered respondents, MTurk subjects report voting at about the same rate as the ANESP, 89 versus 90

⁷ Prospective respondents were offered \$10 per month to complete surveys on the Internet for 30 minutes each month.

⁸ The ANESP redacts age.

percent. They are also slightly more Democratic in their partisan identification than ANESP respondents, and are substantially more extreme on ideology (Table 3).

We also asked the MTurk sample several attitudinal questions that mirrored questions on the ANESP and the ANES (Table 4). These questions asked about support for the prescription drug benefit for seniors, universal healthcare, and a citizenship process for illegal immigrants. The MTurk responses match the ANES well on universal health care – about 50 percent of both samples support it – while those in the ANESP are somewhat less supportive at 42 percent. MTurk also compares reasonably well on the question about a citizenship process for illegal immigrants. Perhaps as a function of the age skew of the sample or a different political environment after the political discussions surrounding the Obama health care initiative, MTurk respondents were less supportive of the prescription drug benefit for seniors compared to the ANES and ANESP—64 percent of MTurk respondents favored the benefit, compared to 75 percent of ANESP and 80 percent of ANES respondents.

Our MTurk survey also included three additional policy questions on the ANESP that were not included on the 2008 ANES. These asked about support for a constitutional amendment banning gay marriage, raising taxes on people making more than \$200,000, and raising taxes on people making less than \$200,000. Compared to the ANESP, MTurk subjects express somewhat more liberal views on all three items, with only 16 percent supporting a constitutional amendment banning gay marriage, compared to 31 percent in the ANESP (Table 4).⁹ On both tax increase items, MTurk subjects are only a few percentage points more liberal in their views.

We also compared political interest and knowledge across the different samples. Individuals likely visit MTurk to make money, not because they are interested in politics. However, we found that MTurk respondents were somewhat more interested than ANESP

⁹ As with the drug benefit, this difference may be due to age or differences in political circumstances.

respondents, and both samples were considerably more interested than ANES respondents (Table 5). These differences may be due to advertising the survey as about "public affairs." We also administered a battery of six political knowledge items from the ANESP. This battery includes questions about the line of succession for the presidency, the length of a U.S. Senate term, and the number of federal senators per state. Just before asking these questions, we instructed respondents to provide just their best guess and not to look up answers. For each item, we offered four answer options in a multiple-choice format. Given this format, only about a quarter should answer each question correctly if randomly guessing. Based on their responses, MTurk subjects appear more knowledgeable than ANESP respondents, but the gap is not large (Table 5). Averaging across the six items common to both surveys, MTurk subjects correctly answer about 71 percent while ANESP respondents get 64 percent right.¹⁰

All told, these comparisons reinforce the conclusion that the MTurk sample does not perfectly match the demographic and attitudinal characteristics of the U.S. population, but also does not present a wildly distorted view of the U.S. population either. It will often be more diverse than convenience samples and always more diverse than student samples. If we treat the MTurk sample as a vehicle for drawing internally valid inferences , instead of a representative sample, the MTurk respondent pool is very attractive.

Benchmarking Via Replication of Experimental Effects

To further assess MTurk's usefulness as a vehicle for experimental research, we also attempted to replicate the results reported in two well-known social science experiments. The

¹⁰ To check whether MTurk subjects looked up answers to knowledge questions on the Internet, we asked two additional questions of much greater difficulty: who was the first Catholic to be a major party candidate for president and who was Woodrow Wilson's vice president. Without cheating, we expected respondents to do no better than chance. On the question about the first Catholic candidate, MTurk subjects did worse than chance with only 10 percent answering correctly (Alfred Smith, many chose an obvious but wrong answer, John F. Kennedy). About a quarter did correctly answer the vice presidential question (Thomas Marshall), what one would expect by chance. These results suggest political knowledge is not inflated much by cheating on MT.

first is a classic study on the sensitivity of public opinion to question wording. The second is a framing experiment. In both cases, the experimental results found using the MTurk sample are highly similar to those found in prior research.

Rasinski (1989) reports results from the question wording experiment study that asked representative samples from the General Social Surveys (GSS) whether too much or too little was being spent on either "welfare" or "assistance to the poor." Even though these terms are thought by policy experts to refer to the same policy, the study found important differences in levels of support. While 20 to 25 percent of the respondents in each year said that too little was being spent on "welfare," 63 to 65 percent said that too little was being spent on "assistance to the poor" (Rasinski 1989, 391). The GSS has continued to ask the spending experiment and the gap remains similar over time, ranging from 28 percent to 50 percent, with an average difference of 37 percent (Green and Kern 2010). When we ran the same experiment on MTurk, we found a similarly-sized gap of 38 percentage points. Only 17 percent said too little was being spent on "welfare," while 55 percent said too little was being spent on "assistance to the poor.

On a separate MTurk survey, we also replicated a classic framing experiment – the "Asian Disease Problem" reported in Tversky and Kahneman (1981). In this experiment, all respondents were initially given the following scenario:

"Imagine that your country is preparing for the outbreak of an unusual disease, which is expected to kill 600 people. Two alternative programs to combat the disease have been proposed. Assume that the exact scientific estimate of the consequences of the programs are as follows:"

They were then randomly assigned to one of the two following conditions:

<u>Condition 1, Lives Saved</u>: If Program A is adopted, 200 people will be saved. If Program B is adopted, there is 1/3 probability that 600 people will be saved, and 2/3 probability that no people will be saved.

<u>Condition 2, Lives Lost</u>: If Program A is adopted 400 people will die. If Program B is adopted there is 1/3 probability that nobody will die, and 2/3 probability that 600 people will die.

These scenarios are exactly the same in their description of the expected consequences of each program, but differ in framing. In condition 1, the two reported programs were described in terms of likelihood of positive outcomes, namely the lives saved by the programs. In condition 2, by contrast, the two programs were described in terms of likelihood of negative outcomes – the lives lost by the different options. Kahneman and Tversky report that when the problem was framed in terms of "lives lost," respondents were more likely to pick the certain choice, while when it was frame in terms of "lives saved," respondents were more likely to pick the risky choice. Framing the outcomes in positive terms therefore produced a reversal of participants' revealed preferences for the two programs compared to when it was presented in negative terms. In the original Kahneman and Tversky experiment, 72 percent of respondents picked the certain choice in the "lives saved" condition, as compared to 22 percent who picked the certain choice in the "lives lost" condition. We find a similar pattern among our MTurk sample: 74 percent picked the certain choice in the "lives saved" condition, and 38 percent selected the certain choice in the "lives lost" condition. Combined with other replications of well-known experiments,¹¹ these findings provide further validation for MTurk as an experimental platform.

Do "Repeat Customers" Dominate the MTurk Subject pool?

One concern with any self-selected sample is the potential for "repeat customers." If the same subjects repeatedly take several surveys, there is the potential for cross-experiment stimuli contamination (Transue et al. 2009). To assess the severity of this problem, we asked our

¹¹ Horton, Rand, and Zeckhauser (2010), for instance, replicate several experimental findings in economics. Gabriele Paolacci's Experimental Turk blog (http://experimentalturk.wordpress.com/) has collected reports of successful replications of several canonical experiments from a diverse group of researchers, including the Asian Disease Problem discussed in this section and other examples from Psychology and Behavioral Economics.

respondents how many political surveys they had taken in the last month on MTurk. The mean was 1.7. Thirty nine percent of the respondents took no other survey, while 78 percent took two or fewer surveys.

We also assessed the prevalence of "repeaters" by examining a broad range of experiments run on MTurk. We gathered the unique MTurk ID number for all workers who participated in each of six studies conducted from January 2010 to April 2010.¹² The compensation for these studies ranged from 10 cents to 50 cents, and the N ranged from 200 to 447. Across the six experiments, there were a total of 1301 unique subjects. Of these subjects, 76 percent participated in only one experiment; another 15 percent participated in two experiments. Only one percent of the subjects participated in 5 or 6 experiments. While this set of experiments represents only a small slice of the experiments conducted on MTurk, our findings call illuminate broader trends. Although there are certainly a handful of respondents who participate repeatedly in experiments, repeat subjects do not appear to be a widespread problem.

Finally, we also sought to assess whether a given respondent took our survey more than once. When we put our study into the field, we specified that each HIT (survey) had to be completed by a unique worker. However, an individual could potentially subvert this process by opening multiple MTurk accounts (though this behavior would violate the terms of the MTurk user agreement). They could then take the survey once from each account. Given the relatively low pay rate of our studies and the availability of other paid work, we do not believe our work is likely to encourage such behavior. Nevertheless, we did check to see if multiple responses came from a single IP address. We found that a total of 7 IP addresses produced two responses each (i.e., 14 of 587 responses or 2.4 percent of the total responses). This pattern is not necessarily evidence of repeat survey taking. It could, for example, be the case that these IP addresses were

¹² This set of experiments did not include the March survey described above.

assigned dynamically to different users at different points in time or that multiple people took the survey from the same large company. But even if these are cases of repeat survey taking, only a handful of responses would be contaminated, suggesting that repeat survey taking is not a large problem in the MTurk system.¹³

Attention and Demand

Given their incentives, MTurk respondents may generally pay greater attention to experimental instruments and survey questions than do other subjects. Since Requesters often specify at least a 95 percent prior "approval rate" – that is, previous Requesters accepted 95 percent or more of the HITs submitted by an individual – respondents have an incentive to read instructions carefully and consider their responses.

Our experiences are consistent with this expectation. In a study conducted by one of the authors, subjects were asked to identify the political office held by a person mentioned in a story they had just read. The format of this question was a multiple choice item with five possible responses. On the Mechanical Turk study, 60 percent of the respondents answered the question correctly. An identical question concerning the same article was also included on experiments run through Polimetrix/YouGov, another high-quality internet panel, and with a sample collected by Survey Sampling International (SSI). The correct answer rates on these platforms were markedly lower than in the Mechanical Turk sample – 49 percent on Polimetrix/YouGov and 46 percent on SSI.

While a concern for pleasing the researcher has benefits, it may also have some costs. MTurk respondents may pay close attention to experimental stimuli, but they may also exhibit experimental demand characteristics to a greater degree than do respondents in other subject

¹³ Researchers can reject and block future work by suspected retakers, or simply exclude duplicate work from their analysis.

pools, divining the experimenter's intent and behaving accordingly (Orne 1962). To avoid this problem and the resulting external validity concerns, we believe subjects should be kept naïve with respect to the purpose of the experiment. Demand concerns are relevant to any experimental research, but future work needs to be done to explore if these concerns are especially serious with respect to the MTurk respondent pool and how they are affected by recruitment and consent text.

Practical Advice: Maximizing Recruitment, Cost Estimates, and Using MTurk to Conduct Panel Surveys

The preceding section considers the strengths and weaknesses of using MTurk to recruit survey experiment subjects. Here we provide advice about how to maximize returns from this apparatus, focusing on two areas of particular interest to researchers: (1) Maximizing recruitment rates for a given task and the associated tradeoffs associated with different pay rates for respondents, and (2) using MTurk to conduct panel surveys.

Compensation and Recruitment

MTurk workers can choose from among many possible jobs, and so descriptions of tasks as "easy," "fun," or otherwise enjoyable and inviting may be likely to increase the rate at which tasks are undertaken. MTurk workers are also being paid, and so promises of prompt payment (including "auto-approving" non-rejected work) are also believed to increase the attractiveness of a job. We know of no research that systematically examines differences in samples produced when using different task descriptions.

Pay relative to length of task is also an important issue for researchers interested in using MTurk. We have tried a variety of pay rates and display examples in Table 6. Although more

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research is needed, we offer a few tentative conclusions. Not surprisingly, paying more per task appears to substantially speed the rate of data collection. For example, when we paid \$.75 for a 5-9 minute survey, 400 respondents took the survey within 24 hours. By contrast, when we paid only \$.25 for a shorter survey (2-3 minutes), 500 respondents took the survey over seven days (see Table 6). We have also noticed a great deal of day-to-day variation in response rates, which may be due to variability in the attractiveness of alternative tasks.¹⁴ Pay and the quality of responses is another important issue. Consistent with our experience, Mason and Watts (2009) find that quality of work on MTurk is not affected by price, but that output declines when wages are lowered.¹⁵

Relative to other experimental pools, MTurk is cheap. Even the higher pay rates we have used on MTurk (effective hourly rates of \$6.00) still are associated with a per-respondent cost of \$.50 for a 5 minute survey or \$.10 per survey minute. By contrast, typical undergraduate (not students in class) cost about \$5-10 per subject and temporary agency subjects cost between \$15 and \$20 per subject. Private survey firms we have worked with charge at least \$10 per subject for a 5 minute survey. MTurk is, in short, extremely inexpensive relative to nearly every alternative other than uncompensated students.

MTurk as a Tool for Panel Research

Another promise of MTurk is as an inexpensive tool for conducting panel studies. Panels studies offer several potential advantages. For example, recent research in political science on the rate at which treatment effects decay (Chong and Druckman 2008; Gerber, Gimpel, Green, Shaw 2007) has led to concerns that survey experiments may overstate the effects of

¹⁴ The website mturk-tracker.com tracks daily data on MTurk. The average daily awards available to workers in 2009 and early 2010 is about \$2,500, but the maximum was almost \$360,000.

¹⁵ Mason and Watts assessed the quality of work in two tasks: sorting images by a timestamp in the images and solving word puzzles.

manipulations relative to what one would observe over longer periods of time. For this reason, scholars are interested in mechanisms for exposing respondents to experimental manipulations and then measuring treatment effects over the long term. Panels also allow researchers to conduct pre-treatment surveys and then administer a treatment distant from that initial measurement (allowing time to serve as substitute for a distracter task). Another potential use of panel study is to screen a large population, and then to select from that initial pool of respondents a subset who better match desired sample characteristics.

The MTurk interface provides a mechanism for performing these sorts of panel studies. To conduct a panel survey, the researcher first fields a task as described above. Next, the researcher posts a new task on the MTurk workspace. We recommend that this task be clearly labeled as open only to prior research participants. Finally, the researcher notifies those workers she wishes to perform the new task of its availability. We have written and tested a customizable PERL script (URL) that does just this (see Appendix 2). In particular, after it is edited to work with the researcher's MTurk account and to describe the new task, it interacts with the Amazon.com Application Program Interface (API) to send messages through the MTurk interface to each invited worker. As with any other task, workers can be directed to an external website and asked to submit a code to receive payment.

Our initial experiences with using MTurk to perform panel studies are positive. In one study, respondents were offered 25 cents for a 3 minute follow-up survey conducted eight days after a first-wave survey. Two reminders were sent. Within five days, 68 percent of the original respondents took the follow up. In a second study, respondents were offered 50 cents for a three minute follow-up survey conducted one to three months after a first-wave interview. Within eight days, almost 60 percent of the original respondents took the follow up.

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findings, Buhrmester, Kwang, and Gosling (In press) report a two-wave panel study, conducted three weeks apart, also achieving a 60 percent response rate. They paid respondents 50 cents for the first wave and 50 cents for the second. Analysis of our two studies suggests that the demographic profile does not change significantly in the follow up survey. Based on these results, we see no obstacle to oversampling demographic or other groups in follow-up surveys, which could allow researchers to study specific groups or improve the representativeness of samples.

Conclusion

This paper describes the potential advantages and limitations of using Amazon.com's Mechanical Turk platform as a subject recruitment device for experimental research. We demonstrate that relative to other convenience samples often used in experimental research in political science, MTurk subjects are generally more representative of the general population and substantially less expensive to recruit. While they are less representative of the population that higher cost internet panels or randomly sampled face-to-face surveys, they appear to respond to canonical experimental stimuli in a manner consistent with prior research. MTurk subjects are apparently also not an excessively sampled pool, and repeat responding appears to be a minor concern. Put simply, despite possible self-selection concerns, the Mechanical Turk subject pool is no worse than convenience samples used by other researchers in political science.

In addition to these benchmarking exercises, we also describe our experiences with maximizing subject recruitment, provide cost estimates, and describe a shared tool we have developed for conducting panel surveys using MTurk. Overall, we believe that these results should provide researchers with a clearer understandings of the potential advantages of the MTurk tool for conducting experiments as well as areas where caution may be in order.

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		Conven	ience Samj	•	Face-To Samp	
		Student sample (Kam et al	Adult samples (Berinsky and Kinder 2006)		CPS	ANES '08
Demographics	MTurk	2007)	Exp. 1	Exp. 2		
Female	60.1%	56.7%	66.0%	57.1%	51.7%	54.0%
	(2.1)	(1.3)			(0.2)	(1.3)
Age (mean years)	32.3	20.3	42.5	45.3	46.0	46.6
	(0.5)	(8.2)			(0.1)	(0.5)
Education (mean years)	14.9		15.1	14.9	13.2	13.5
	(0.1)				(0.0)	(0.1)
White	83.5	42.5	81.4	72.4	81.2	79.1
	(1.6)				(0.1)	(0.9)
Black	4.4		12.9	22.7	11.8	12.0
	(0.9)				(0.1)	(0.6)
Party Identification						
Democrat	41.9		46.1	46.5		34.2
	(0.8)					(1.2)
Independent	23.1		20.6	17.6		28.7
	(0.6)					(1.2)
Republican	24.9		16.3	25.8		25.9
_	(0.7)					(1.2)
None/other	10.2		17.0	10.1		11.3
	(0.5)					(0.8)

Table 1: Benchmarking MTurk against Convenience Samples

Note: Percentages except for age and education with standard errors in parentheses. CPS and ANES '08 are weighted.

		Internet Sample	Face-To-Face Samples				
	MTurk	ANESPS	CPS	ANES '08			
Female	60.1%	57.6%	51.7%	54.0%			
	(2.1)	(0.9)	(0.2)	(1.3)			
Age (mean years)	32.3		46.0	46.6			
	(0.5)		(0.1)	(0.5)			
Education (mean years)	14.9	14.5	13.2	13.5			
	(0.1)	(0.1)	(0.0)	(0.1)			
Race							
White	83.5	84.3	81.2	79.1			
	(1.6)	(0.7)	(0.1)	(0.9)			
Black	4.4	9.0	11.8	12.0			
	(0.9)	(0.5)	(0.1)	(0.6)			
Hispanic	6.7	5.5	13.7	9.1			
	(1.1)	(0.4)	(0.1)	(0.5)			
Marital Status							
Married	39.0	63.6	55.7	50.1			
	(2.1)	(0.9)	(0.2)	(1.3)			
Divorced	7.1	13.6	10.2	12.9			
	(1.1)	(0.7)	(0.1)	(0.8)			
Separated	2.5	1.5	2.1	2.9			
I	(0.7)	(0.2)	(0.1)	(0.4)			
Never married	50.6	15.9	25.7	24.7			
	(2.1)	(0.7)	(0.2)	(1.1)			
Widowed	0.7	5.4	6.3	7.8			
	(0.4)	(0.4)	(0.1)	(0.6)			
Housing Status	(011)	(011)	(011)	(0.0)			
Rent	52.7	14.3		32.4			
	(2.3)	(0.7)		(1.2)			
Own home	47.3	80.8		66.1			
	(2.3)	(0.8)		(1.2)			
Religion	(2.5)	(0.0)		(1.2)			
None	41.8	13.3		20.1			
Tone	(2.1)	(1.0)		(1.1)			
Protestant	20.7	36.4		30.6			
1 IOUStallt	(1.7)	(1.4)		(1.2)			
Catholic	16.5	24.2		(1.2)			
Catholic	(1.6)			(1.0)			
Jewish	(1.6)	(1.3) 3.8		(1.0)			
JEWISII							
Other	(0.9)	(0.6)		(0.3)			
Other	16.5	22.3		28.9			
	(1.6)	(1.2)		(1.1)			

Table 2: Benchmarking MTurk on Demographics

Note: Percentages except for age and education with standard errors in parentheses. CPS and ANES '08 are weighted. ANESP redacts age.

Table 5. Determarking M	Internet Face-To-Face S				
		Sample	CDC		
Desistantion and Typesout	MTurk	ANESPS	CPS	ANES '08	
Registration and Turnout	79.70/	02.00/	977 0/	77 70/	
Registered	78.7%	92.0%	82.7%	77.7%	
Mate Turn out	(1.7)	(0.6)	(0.2)	(1.4)	
Vote Turnout	88.6 (1.5)	89.5 (0.6)	74.5 (0.2)	77.6	
Party Identification	(1.3)	(0.0)	(0.2)	(1.1)	
Strong Democrat	17.8	20.4		20.3	
Strong Democrat	(1.7)	(1.0)		(1.0)	
Weak Democrat	26.8	14.9		16.4	
Weak Democrat	(2.0)	(0.9)		(1.0)	
Lean Democrat	7.7	10.7		16.1	
	(1.2)	(0.8)		(1.0)	
Independent	12.4	12.1		8.5	
	(1.5)	(0.8)		(0.8)	
Lean Republican	17.0	9.2		10.9	
	(1.7)	(0.7)		(0.9)	
Weak Republican	9.1	15.5		13.7	
I I I I I I I I I I I I I I I I I I I	(1.3)	(0.9)		(1.0)	
Strong Republican	9.3	17.2		14.1	
0 1	(1.3)	(0.9)		(1.0)	
Ideology		. ,		. ,	
Very liberal	26.2	7.2		3.7	
	(1.9)	(0.6)		(0.6)	
Somewhat liberal	21.7	14.8		13.0.	
	(1.8)	(0.9)		(1.0)	
Slightly liberal	5.1	11.2		11.8	
	(0.9)	(0.8)		(1.0)	
Moderate	17.5	22.0		29.1	
	(1.6)	(1.0)		(1.4)	
Slightly conservative	7.9	8.7		16.0	
	(1.1)	(0.7)		(1.1)	
Somewhat conservative	6.7	23.9		22.4	
	(1.1)	(1.1)		(1.3)	
Very conservative	15.0	12.3		4.1	
	(1.5)	(0.8)		(0.6)	

Table 3: Benchmarking MTurk on Voter Turnout, Party, and Ideology

Note: Percentages with standard errors in parentheses. CPS and ANES '08 are weighted.

	MTurk	Internet Sample ANESPS	Face-To- Face Samples ANES '08
Favor prescription drug benefit for seniors	63.5%	74.8%	80.1%
	(2.0)	(1.1)	(1.5)
Favor universal healthcare	47.8	41.7	51.0
	(2.1)	(1.2)	(1.9)
Favor citizenship process for illegals	38.4	42.7	49.1
	(2.1)	(1.2)	(1.9)
Favor a constitutional amendment banning gay marriage	15.5	30.7	
	(1.5)	(1.2)	
Favor raising taxes on people making more than \$200,000	61.4	55.4	
	(2.1)	(1.2)	
Favor raising tax on people making less than \$200,000	6.1	7.1	
	(0.6)	(0.6)	

Table 4: Benchmarking MTurk on Policy

Note: Percentages with standard errors in parentheses. ANES '08 is weighted.

	MTurk	Internet Sample ANESPS	Face-To- Face Samples ANES '08
Political interest			
Interest: Extremely	19.7%	11.8%	8.3%
	(1.7)	(0.6)	(1.0)
Interest: Very	34.1	31.6	23.5
	(2.0)	(0.9)	(1.6)
Interest: Moderately	32.5	35.9	41.4
	(2.0)	(0.9)	(1.9)
Interest: Slightly	10.6	15.0	19.5
	(1.3)	(0.7)	(1.5)
Interest: Not at all	3.1	5.7	7.2
	(0.7)	(0.4)	(0.9)
Political knowledge			
Presidential succession after Vice President	70.0	65.2	
	(1.3)	(2.0)	
House vote percentage needed to override a veto	81.3	73.6	
	(1.7)	(1.3)	
Number of terms to which an individual can be elected president	96.2	92.8	
1	(0.8)	(0.7)	
Length of a U.S. Senate term	45.0	37.5	
	(2.1)	(1.3)	
Number of Senators per State	85.4	73.2	
L	(1.5)	(1.2)	
Length of a U.S. House term	50.1	38.9	
č	(2.1)	(1.3)	
Average	71.3	63.5	

Table 5: Benchmarking MTurk on Political Interest and Political Knowledge

Note: Percentages with standard errors in parentheses. ANES '08 is weighted.

Study Title	Date Launched	Ν	Pay	Total Cost	Mean Mins. on Job	1	2	3	4	5	6	7
2-3 minute survey for Yale political science research	3/16/2010	500	\$0.25	\$57.75	2	210	68	37	55	53	64	18
4 minute survey for Yale political science research	4/26/2010	500	\$0.40	\$119.20	4	298	105	79	18			
3 minute survey for Yale political science research	4/29/2010	200	\$0.25	\$50.00	1	200						
3-4 minute survey for Yale political science research	5/17/2010	150	\$0.45	\$67.50	2	150						
5-7 minute survey	6/28/2010	400	\$0.75	\$240.75	5	321	79					
7-9 minute survey	6/24/2010	400	\$0.75	\$300.00	6	400						
5-7 minute survey	7/3/2010	400	\$0.50	\$128.00	3	256	115	29				
2-3 minute survey	7/16/2010	200	\$0.25	\$50.00	3	200						

Table 6: Examples of Response Rates to Mechanical Turk Studies