

Who Cooperates? Reciprocity and the Causal Effect of Expected Cooperation in Representative Samples

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Abstract

When do societies succeed in providing public goods? Previous research suggests that public goods contributions correlate with expectations about cooperation by others among students and other demographic subgroups. However, we lack knowledge about whether the effect of expected cooperation is causal and a general feature of populations. We fielded representative surveys ($N = 8,500$) in France, Germany, the United Kingdom, and the United States that included a public goods game and a novel between-subjects experiment. The experiment varied expectations about cooperation by others. We find that higher expected cooperation by others causes a significant increase in individual contributions. When classifying contribution schedules, we find that almost 50% of the population employs a conditionally cooperative strategy. These individuals are on average richer, younger, and more educated. Our results help explain the varying success of societal groups in overcoming cooperation problems and assist policymakers in the design of institutions meant to solve social dilemmas.

Keywords: Public goods, social norms, conditional cooperation, reciprocity, survey experiment, representative samples

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Societies face numerous public goods problems and their solution requires cooperation. Examples cover a wide range of policy areas such as global warming (Barrett, 2003), economic crises and poverty (Kaul et al., 1999), international security (Sandler and Hartley, 1999), regime change (Hollyer et al., 2015), public health (Smith et al., 2003), and knowledge (Stiglitz, 1999). What explains the varying success of societies in realizing cooperation in the face of collective action problems? A large set of literatures in political science and many other academic disciplines have addressed this question both theoretically and empirically (Hume, 2003; Olson, 1965; Axelrod, 1984; Ward, 1990; Keohane and Ostrom, 1995). The cumulative empirical insights of lab and lab-in-the-field experiments investigating why groups succeed or fail in providing public goods have been central to this body of research. These studies strongly suggest that reciprocity, or the willingness to cooperate if one expects others to do so, constitutes a particularly influential determinant of actors' behavior in public goods games when examining students (Fischbacher and Gächter, 2010; Engelmann and Strobel, 2010; Charness and Haruvy, 2002; Ostrom, 2000) and other selected demographic subgroups (Fehr and Fischbacher, 2002; Gächter et al., 2004; Henrich et al., 2001).

This work has also inspired a growing literature in political science that highlights the importance of conditional cooperation for crafting political solutions to collective action problems. Examples include climate change (Bechtel et al., 2017; Tingley and Tomz, 2014; Bechtel and Scheve, 2013), tax policy (Frey and Torgler, 2007), political participation (Bolsen et al., 2014), and ethnic conflict (Habyarimana et al., 2007). These studies have in common that conditional cooperation appears to play an important role for our understanding of when societies succeed in solving collective action problems. Recent micro-level research has even begun to identify different types of conditionally cooperative individuals to explain variation in public goods provision over time (Fischbacher et al., 2001; Fischbacher and Gächter, 2010) and in ethnically diverse societies (Habyarimana et al., 2007). This is because the exact functional form that individuals use to map from other contributions to their own contribution in public goods games may vary greatly and this variation in strategy types can have profound consequences for both the possibility of cooperation and the role of institutions in promoting the provision of public goods (Ostrom, 2000).

Although much has been learned from this work, a widely shared concern about these results has been that they are largely based on student populations. Do more diverse and representative samples of subjects behave differently in social dilemmas? This question mirrors similar concerns that have been articulated in the case of online labor markets for political science experiments (Huff and Tingley, 2015; Berinsky et al., 2012) or regarding the samples used in education research that may not be representative of the population that researchers are actually interested in (Kern et al., 2016).

While examining students or other selected subgroups of the adult population will yield generalizable results under the assumption of homogeneous treatment

effects (Druckman and Kam, 2011), the plausibility of this assumption needs to be assessed empirically. Scholarship has begun to address this question primarily by moving labs to the field and examining behavior in ultimatum and trust games in more diverse, nonstudent samples (Gächter et al., 2004; Bellemare and Kröger, 2007; Holm and Danielson, 2005; Tsai, 2007; Habyarimana et al., 2007). Yet, the subject pools are still, with a few exceptions (Fehr et al., 2002; Thöni et al., 2012), based on selected samples that are not nationally representative. More importantly, the results from these studies suggest that our ability to generalize from students to other sub-groups of the population may be limited. For example, Japanese fishermen (Stoop et al., 2012), adult trainee truckers (Anderson et al., 2013), as well as adults from Oxford (Belot et al., 2015) exhibit more cooperative behavior than students when making decisions in a laboratory environment. In fact, Belot et al. (2015) find that pro-social behavior differs most strongly when comparing students and nonstudent populations. Therefore, we still lack knowledge about public goods contributions and conditional cooperation among the general population in industrialized democracies. Filling this gap in the literature requires the use of nationally representative samples.

This paper provides the first evidence on public goods contributions in representative samples for four industrialized democracies and implements a novel between-subjects design for estimating the causal effect of expected cooperation. We fielded a large-scale survey ($N = 8,500$) in France, Germany, the United Kingdom, and the United States that included a two-player, simultaneous, anonymous public goods game. Our design uses both correlational variation within subjects and experimental variation between subjects to explore the effects of expected cooperation on individuals' contributions to a public good. The causal evidence we present is based on a novel randomized experiment in the game instructions to respondents and an instrument that elicited the individual's conditional contribution schedules (Rauhut and Winter, 2010; Fischbacher et al., 2001; Selten, 1967). The experiment randomized the example with which the instructions illustrated how the ultimate pay-off depends on both players' actions. Thus, our version of an encouragement design experimentally manipulated whether respondents believed they were acting in a more or less cooperative environment. It provides between-subjects causal evidence that expectations about the cooperative behavior of others influence public goods contributions.

We find that, while socio-demographics are at most weakly correlated with individuals' cooperative behavior, expectations about contributions by the other player are strong predictors of cooperation. Based on our randomized experiment, we provide the first between-subjects estimate of the causal effect of expected cooperation on public goods contributions in representative samples. We find that a 1 €/£/\$ increase in the contribution an individual expects from the other respondent causes 1.4 €/£/\$ higher own contributions on average. While this effect size suggests that individuals over-reciprocate, we

find that a 95% confidence interval for this estimate contains 1. This result significantly strengthens the empirical literature on conditional cooperation: The effect of expected cooperation on contribution behavior is substantively large, clearly evident in a representative set of subjects, and can be given a causal interpretation.

In addition to studying the average effects of conditional cooperation, we investigate who employs reciprocal strategies and how these reciprocity types condition the causal effect of expected cooperation. We classify our 8,500 respondents into different groups of conditional cooperators (Fischbacher et al., 2001; Thöni et al., 2012), distinguishing between Positive Reciprocity (44%), Positive Nonconditional (13%), Freerider (11%), Inverse U-shaped Reciprocity (5%), and Other (27%). Although the distribution of these types is quite similar across our four countries, we find that they are not evenly spread throughout the populations. Specifically, we find that positive reciprocity types, i.e., individuals that contribute more if others contribute more, are significantly more widespread among female, younger, wealthier, and more educated individuals. Further, we find that even among positive reciprocity types, younger, wealthier, and highly educated individuals respond more strongly to their expectations of the contributions of others to the public good. Finally, we re-estimate the causal effect of cooperative expectations for different reciprocity types using the experimental design described above. We find that the effects of expectations about the contributions of others are strongest among positive reciprocity types and generally insignificant for most of the other strategy types.

Taken together, these results provide new and compelling evidence of the important causal role that reciprocity plays in the provision of public goods. Further, our analysis suggests that the ability of researchers to generalize from student samples to other subgroups depends on the research question and the target population of interest. Our results indicate that the overall level and distribution of contributions in public goods games are similar in representative and student samples. However, we also find important differences in the extent and type of conditional cooperation across different socio-demographic groups. Individuals that employ conditionally cooperative strategies are disproportionately younger, wealthier, and more educated. We also show that the causal effect of expected cooperation is greater among individuals employing conditionally cooperative strategies. These socio-demographic characteristics vary either slowly over time or are largely time-invariant. Since previous research has suggested that social preferences exhibit a high degree of stability over the long run (Carlsson et al., 2014), this heterogeneity could be substantively consequential for understanding cooperation across groups and helps to shed light on when researchers may find employing more representative albeit more costly samples useful. We discuss the implications of these findings for our understanding of cooperation in social dilemmas, the role of policy interventions and institutions for equilibrium selection in collective action problems, and subsequent research on reciprocity.

SURVEY, SAMPLE, AND EXPERIMENTAL DESIGN

Our survey was conducted on representative samples of the adult population in France ($N = 2,000$), Germany ($N = 2,000$), the United Kingdom ($N = 2,000$), and the United States ($N = 2,500$). The surveys were carried out online by YouGov in summer 2012. YouGov employs matched sampling to approximate a random sample of the adult population (Rivers, 2011). The appendix provides details about the survey design including marginal distributions of socio-demographics in the sample and the populations (see Table A1). These descriptives reveal that the socio-demographic margins of the samples in the four countries are very similar to that of the target populations. However, since matching is approximate, survey weights were computed based on official information about the distribution of socio-demographic characteristics to eliminate the remaining imbalances. The results we report below are based on weighted data but remain virtually unchanged when reestimated without weights.

The survey had two parts. The first part was an anonymous, two-player public goods game that was embedded in a randomized experiment. This component of the survey provides us with exogenous between-subjects variation in individuals' levels of expected cooperation. The experiment did not involve deception. The second part elicited respondents' contribution schedules using the strategy method (Selten, 1967) that asks individuals to indicate how much they are willing to contribute given the other player's contribution.

Public goods game and other contribution experiment

Our survey informed all respondents that they could win one of two Amazon gift cards and that the amount of the gift card would depend (i) on their decision about whether to give some amount of the gift card to the other winner and (ii) the analogous decision made by that winning respondent. The survey left the probability of winning a gift card unspecified.¹ Any amount given to the other respondent would be subtracted from the individual's base winnings of 100 €/£/\$ and doubled before it was distributed to the other winner. Thus, the public good in this case equals the amount of money that the group takes from the experimenter. Formally, the payoff function for an individual i is: $\Pi_i = 100 - c_i + 2c_j$, where c_i denotes one's own contribution and c_j is the contribution by the other individual. Standard economic theory predicts that individuals should contribute nothing, which implies that the payout to the two winners is the minimum aggregate payoff of 200 €/£/\$. However, the highest possible aggregate payoff is 400 €/£/\$. The exact instructions that followed the description of the raffles were the following:

¹The true probability was 0.1% in France, Germany, and the United Kingdom and 0.08% in the United States. In the public goods game, the gift card winners received \$140 on average. Independent of their answers, all respondents also received a fixed incentive for taking part in the survey administered by YouGov.

The ultimate value of the voucher depends on your decision on the following: If you win a voucher, you can decide to increase the value of the second voucher that another person has won. You can give any amount between 0 and 100 €/£/\$ by which the value of your voucher will be decreased. Each €/£/\$ that you decide to give to the other individual will be doubled. This means that if you decide to give, say, [10, 30, 60, 90]€/£/\$, the other person will receive (20, 60, 120, 180) €/£/\$ and you will receive (90, 70, 40, 10)€/£/\$. Likewise, if the other person decides to give, say, [10, 30, 60, 90]€/£/\$, you will receive (20, 60, 120, 180)€/£/\$ and they will receive (90, 70, 40, 10)€/£/\$.

We randomized the value in square brackets and computed the corresponding values stated in parentheses. As an example, one possible realization of the randomized part of the instructions was: “This means that if you decide to give, say, \$10, the other person will receive \$20 and you will receive \$90. Likewise, if the other person decides to give, say, \$60, you will receive \$120 and they will receive \$40.” Thereby, we exogenously varied whether respondents were acting in a more or less cooperative environment. The design of our other contribution experiment therefore resembles a mild version of an encouragement design that typically invites subjects randomly to participate in a treatment condition if a randomized control trial does not seem feasible because of ethical or practical reasons (West et al., 2008).² In our application, informing subjects that the other winner would contribute a specific amount would be dishonest (Morton and Williams, 2010; Dickson, 2011). Our version of an encouragement design that merely provides examples included in the instructions to explain the payoff mechanism enables us to avoid deception.

We subsequently asked respondents how much they would like to contribute and how much they expect the other winner to contribute (the order in which we asked these questions was randomized). After completion of the field work, the winners were drawn and their contribution decisions determined the payoffs. When comparing our results with those reported in lab-experimental studies with selected subsets of the population, it may be relevant to consider the level of incentives. However, a review of more than 70 experiments concludes that differences in incentivization do not affect subjects' contributions on average (Camerer and Hogarth, 1999). Moreover, Gillis and Hettler (2007) show that even hypothetical incentives tend to replicate behavior in public goods games that are incentivized.

Measuring different types of conditional cooperation

The second part of the survey asked respondents about their conditional contribution schedules. To determine individuals' types of conditionally cooperative strategies, immediately after our survey respondents played the payoff-relevant public goods game, we asked them to indicate how their own contribution

²There exist several possible mechanisms that could bring about the variation in individual expectations about what the other person will contribute, for example, anchoring or demand effects (Roux and Thöni, 2015). We remain agnostic about the exact mechanism as our primary purpose is to create exogenous variation in respondents' expectations about the other individual's contribution.

potentially depends on the other individual's contribution (Rauhut and Winter, 2010; Fischbacher et al., 2001; Selten, 1967). The exact question wording was the following:

Now suppose you knew how much the other winner of the voucher was going to contribute. Please indicate how much you would like to give if the other winner of the voucher gives the following amount? Remember that any amount that you decide to give to the other winner is doubled.

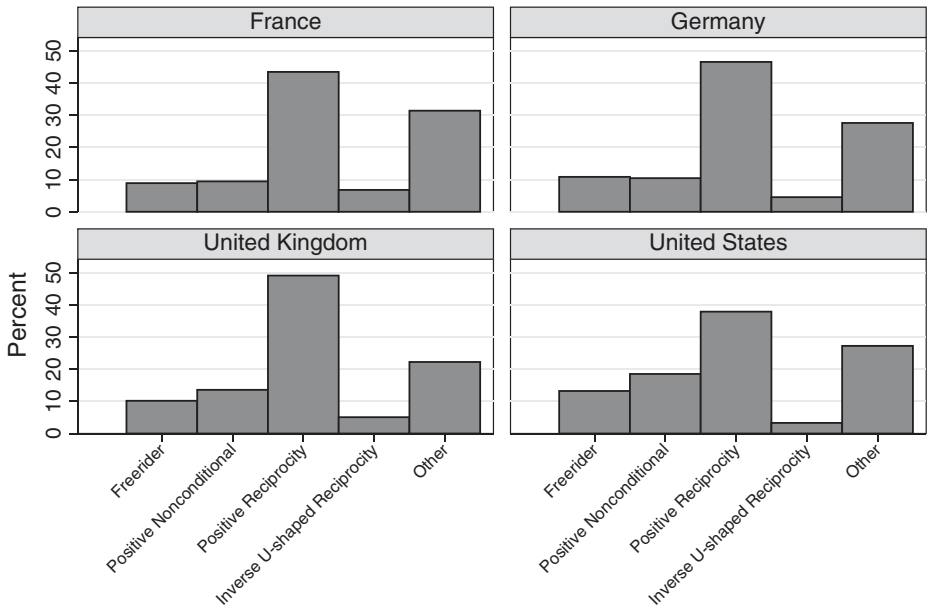
The respondent then chose a value they would give if they knew the other winner contributed 0, 25, 50, 75, 100 €/£/\$.

Figure A1 shows the 8,500 individual contribution mappings with darker lines indicating higher frequency. This part of the survey was not explicitly incentivized. To probe whether this part of the survey generated differences in individuals' level of conditional cooperation, we used an individual's expectation about the other contribution and his/her strategy (which maps from other to own contribution) to generate the contribution we would expect him/her to make based on his/her answers in the strategy method part of the survey, which asked respondents to indicate how much they would contribute given the other player's contribution. The appendix provides more details about this additional analysis and the results. We find a very strong positive correlation between individual's contribution choices in the payoff-relevant part of the survey and the strategy method part. This suggests that the answers in these two parts of the survey are consistent. The results are reported in Table A6 (see also footnote 7).

To explore the distribution of strategy types we code respondents depending on the functional form that relates their own contribution and the contribution by the other individual (the appendix provides detailed coding rules). Our approach employs a computational algorithm in the spirit of Kurzban and Houser (2005) and uses a slightly more fine-grained classification than previous work (Fischbacher et al., 2001; Thöni et al., 2012) to distinguish five types: Freerider, Positive Nonconditional, Positive Reciprocity, Inverse U-shaped Reciprocity, and Other. We add the category of "Positive Nonconditional" individuals because we empirically observe a nonnegligible share of respondents (about 10% or 800 individuals) that make constant, positive contributions. Those individuals do not plausibly exhibit an erratic pattern that one would want to classify as "other." Given that this type of behavior also seems theoretically important, we add it as an additional category.³

Respondents are coded as Freeriders if they always contribute less than 5 €/£/\$. We code an individual as Positive Nonconditional if she/he gives a constant positive contribution that does not vary across the different known values of the other winner's contribution (graph is a horizontal line placed above 0).

³The fact that this type was not observed in Fischbacher et al. (2001) may either be due to their small sample of students (44 subjects) or reflect that students are truly less nonconditionally cooperative on average.



Graphs by country

Figure 1
Distributions of Strategy Types by Country.

This plot shows the distribution of individuals’ strategy types. Strategies are elicited using the strategy method (see section “Coding of Strategies” for details).

The horizontal line need not be perfectly flat but cannot vary across all values by more than 5 \$/£/€. Positive reciprocity types are individuals whose contributions increase monotonically and the total increase is greater than 5 \$/£/€. We also identify inverse U-shaped reciprocity types (sometimes called “triangle contributors,” see Fischbacher and Gächter (2010)) whose contribution function is convex and the difference between the maximum and minimum contribution is greater than 5 \$/£/€. Types that do not fit any of the definitions above are coded as Other. We also reestimated all results using 10 \$/£/€ as the threshold. Our findings remain virtually identical.

Figure 1 shows the distributions of strategy types by country. Across all four countries we find largely similar distributions of individuals’ strategy types. About 10% of the populations are Freeriders and roughly the same share can be classified as Positive Nonconditional. Between 40% and 50% of the societies consist of positive reciprocators. Finally, about one third of our representative sample uses some other functional form to map from their expectations about other contributions to their own contribution level.

How does the distribution of strategy types, which is based on a large, representative sample, compare with those reported in lab-experimental studies

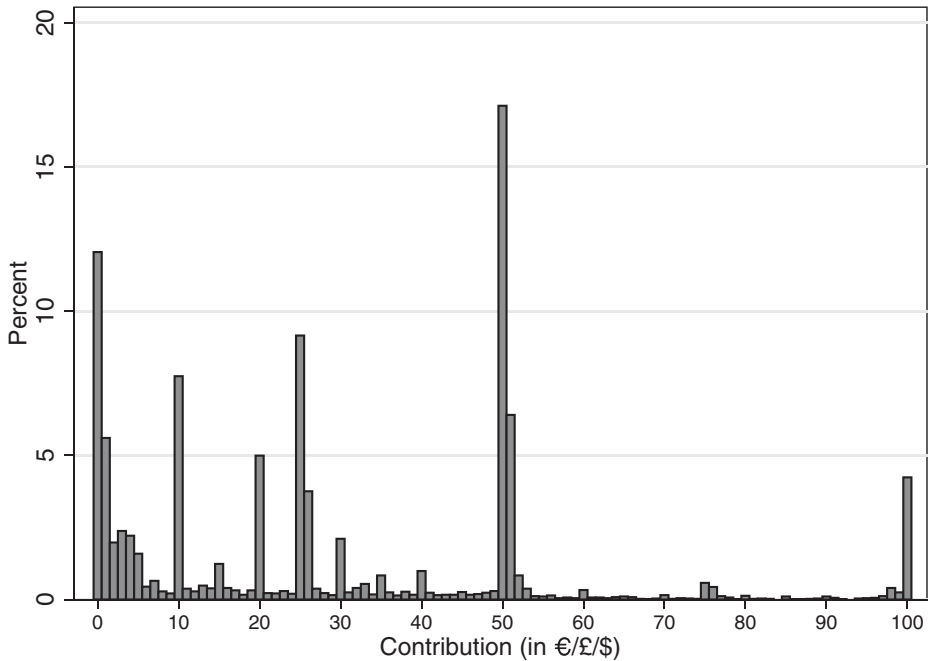


Figure 2

Distribution of Own Contributions to the Public Good in Representative Samples.

The plot shows the distribution of contributions to the public good (pooled data, $N = 8,500$). Countries included are France ($N = 2,000$), Germany ($N = 2,000$), the United Kingdom ($N = 2,000$), and the United States ($N = 2,500$).

with students as subjects? We find that the share of positive reciprocators largely equals that reported in lab-experimental studies (Fischbacher et al., 2001). However, we observe a smaller share of freeriders (10%) than previous work examining students as subjects. For example, Fischbacher et al. (2001) classify 13 out of 44 students (30%) as purely selfish. Also, in our large, representative sample the share of “hump-shaped” contribution schedules is smaller (about 5%) than in their student sample (14% or 6 out of 44).

RESULTS

Own contribution behavior and expected contributions

Figure 2 shows the distribution of individual contributions in the public goods game for the pooled data ($N = 8,500$). Only about 12% of the individuals in France, Germany, the United Kingdom, and the United States make zero contributions. The vast majority contributes substantively to the public good. The modal contribution is 50 €/£/\$. Overall, contributions cluster at 0, 10, 20, 25, 50,

and 100 €/£/\$ and the average contribution is 29.4 €/£/\$. The high percentage of respondents that contribute to the public good resonates with the main claim of the lab-experimental literature that cooperative behavior in public goods games is substantially more prevalent than predicted by standard economic theory. Moreover, the average contribution implies a 29.4% of endowment contribution, which is broadly similar to average values reported in laboratory studies for one-shot static public goods games like ours.⁴

Figure 2 reveals substantial variation in the distribution of public goods contributions in a representative population. What explains this variation and what is the relative importance of socio-demographic characteristics as opposed to norms of conditional cooperation? Since we examine cooperative behavior in representative samples with over 8,000 respondents who offer a lot of variation in socio-demographics attributes, we have the rare opportunity to explore if public goods contributions cluster along socio-demographic cleavages or whether they mostly reflect the presence or absence of reciprocal expectations. Thereby, our study also provides evidence on the generalizability of lab-experimental findings with selected subgroups of populations. To the extent that socio-demographic differences do not exist, studying selected subgroups reduces research costs without compromising on the generalizability of the findings.

We first explore the socio-demographic distribution of cooperation. To this end, we constructed the variable *Own Contribution* equal to each respondent's stated contribution to the linear public goods game. Figure 3 reports the ordinary least squares coefficient estimates—with 95% confidence intervals based on heteroskedasticity-robust standard errors—for the regression of *Own Contribution* on indicator variables for sex, age, income, and education. The results are striking. In general, socio-demographic characteristics are uncorrelated with observed contribution behavior. The estimates reported in Figure 3 are small and statistically insignificant for age, income, and education. The one exception to this pattern is that on average, women tend to contribute about 2 €/£/\$ less than men. This stands in contrast to most lab-experimental studies with students and other selected subgroups of the population, which report that female subjects tend to contribute more than male individuals.⁵ This discrepancy could indicate that female students in lab-experimental studies behave differently than women in the overall population, although other explanations, for example, that women respond more strongly to implicit or explicit cues provided by the experimental setting (Croson and Gneezy, 2009) remain possible. We also found no significant differences in our pooled analyses by marital status, employment status, and ideology.⁶

⁴For example, Fischbacher et al. (2001) report that the total average “unconditional contribution” for a static one-shot public goods game is 33.5% of the initial endowment in their study.

⁵See Croson and Gneezy (2009) for a detailed review of this work.

⁶See Table A2 for additional results including tobit estimations that account for the censoring of contributions at 0 and 100. In country-specific analyses, we also examined whether there were differences by partisan identification. In France, individuals identifying with the National Front on average

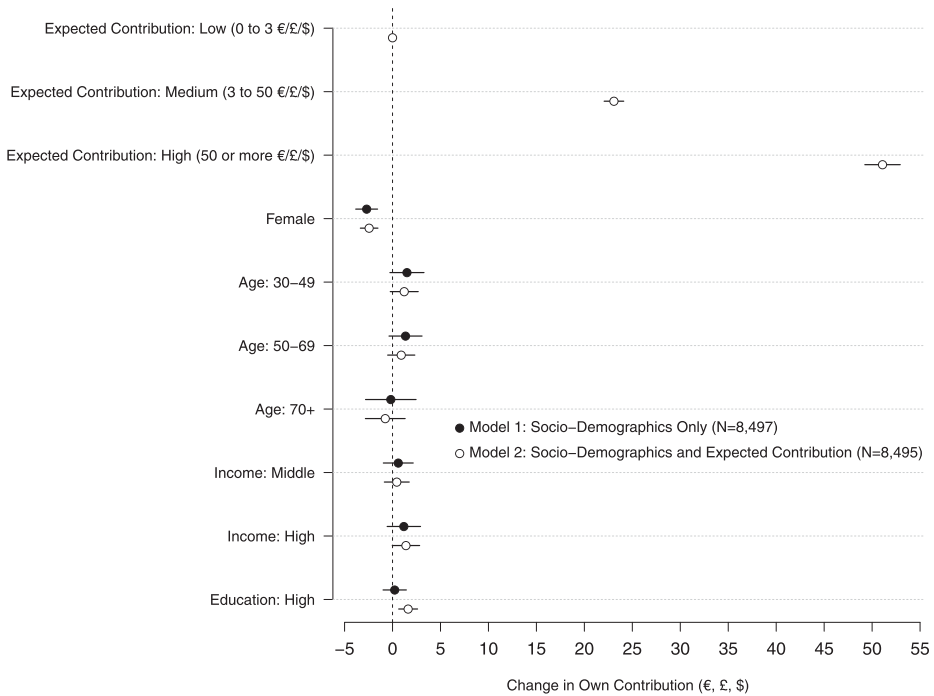


Figure 3

Correlates of Cooperative Behavior in France, Germany, the United Kingdom, and the United States (Pooled Data).

This plot shows OLS coefficients (dots) and 95% confidence intervals computed from heteroskedasticity-robust standard errors (Model 1: $N = 8,497$ and Model 2: $N = 8,495$). In Model 1, we regressed an individual's own contribution on socio-demographic characteristics. In Model 2, we regressed an individual's own contribution to the public good on the socio-demographics and two indicator variables that capture how much a respondent expects the other player to contribute. Expected contribution is measured using the following question: "How much do you think the other winner will contribute?" Answers were coded using three categories based on the distribution of responses. Expected contributions are coded as low if the respondent expects the other winner to contribute below 3 €/£/\$ (25th percentile). Expected contributions are coded as medium if the expected contribution is between 3 €/£/\$ and below or equal to 50 €/£/\$ (75th percentile), and high if the expected contribution is greater than 50 €/£/\$. The coefficient without a confidence interval indicates the reference group (Expected Contribution: Low). All models include country fixed effects.

Overall, there exists little evidence in our data to suggest that some socio-demographic groups cooperate more in our public goods game than others. This evidence helps in assessing the plausibility of explanations for variation in public goods contributions. For example, it is inconsistent with the idea that high income individuals feel better able to afford contributions and thus find it easier to cooperate among themselves. These results are also inconsistent with explanations based on the idea that higher educated individuals have the ability to realize the advantages of everyone contributing and thus cooperate more easily.

contributed 5 € less than those not identifying with a party. In Germany, individuals identifying with the CDU on average contributed 4 € more than those not identifying with a party. We observed no significant partisan differences in the United Kingdom and the United States, though in the United Kingdom right ideology was modestly correlated with lower contributions.

Our primary interest is in determining the importance of conditional cooperation in explaining contribution behavior in our large-scale, representative sample. We converted our continuous measure of the expected contribution of the other respondent, *Expected Contribution*, into three indicator variables based on the 25th and the 75th percentile of the observed distribution and added them to the regression model. **Figure 3** also reports these estimates. Compared to those with a low expected contribution (between 0 and below 3 €/£/\$), respondents that expect a medium contribution (3 to below 50 €/£/\$) decide to contribute about 24 €/£/\$ more on average. Respondents that expect the other winner to make a high contribution (50 or more €/£/\$) contribute about 50 €/£/\$ more on average. We find that, when compared to coefficients on the socio-demographic variables, an individual's expectation about how much the other individual will contribute is, by far, the strongest predictor. This finding remains unchanged if we re-estimate the model using a continuous measure of expectations, *Expected Contribution*, where we obtain a highly significant and positive coefficient (0.73, $p < 0.001$).

In an additional analysis reported in Table A2 (Model 2), we add a large set of covariates including a variable that captures individuals' level of altruism using a quasi-behavioral measure of altruistic inclination. This measure is based on the following survey instrument: We informed respondents that we will raffle another 100 €/£/\$ among all respondents that completed the survey and that the winner can decide to donate parts of the voucher to a charity. We then asked respondents whether they would like to donate in case they won a voucher. If respondents indicated that they wanted to donate, we offered a large list of charities from which individuals could choose and asked them about the amount they would like to give. We coded respondents as exhibiting a relatively high level of altruism if, within the group of those willing to donate, they donated a nonzero amount (the median donation was zero). Consistent with previous laboratory results, we find that more altruistic individuals contribute significantly more in the public goods game, but we also find that the point estimates for the expected cooperation indicators remain largely unchanged even when accounting for altruism and other potential confounds.

We also reestimated the model separately for those randomly selected 50% of respondents that were first asked to indicate their own contribution and for the remaining 50% of respondents that were first asked to indicate their expectations about how much the other winner contributes. The results for these two subgroups are virtually identical (see Models 3 and 4 in Table A2). Overall, the findings from our representative sample are consistent with the view that reciprocity is an important norm that guides individual contribution behavior.

The causal effect of expected cooperation

The partial correlation between expectations about the cooperative behavior of others and an individual's own contribution may or may not reflect a causal

Table 1
The Causal Effect of Expected Contribution on Own Contribution: Reduced Form and Instrumental Variable Results

Outcome Variable	(1) Reduced- form Own Contribution	(2)	(3) IV First- stage Expected Contribution	(4)	(5) IV Own Contribution	(6)
Expected Contribution (IV)					1.38*** (0.372)	1.45*** (0.389)
Other Contribution Treatment: High	2.38*** (0.658)	2.45*** (0.657)	1.73*** (0.649)	1.70*** (0.646)		
Female		-2.73*** (0.581)		-2.02*** (0.567)		0.19 (0.987)
Age: 30–49		1.52* (0.905)		0.49 (0.905)		0.81 (0.950)
Age: 50–69		1.34 (0.876)		0.45 (0.872)		0.70 (0.923)
Age: 70+		-0.12 (1.340)		-0.47 (1.311)		0.56 (1.305)
Income: Middle		0.63 (0.794)		0.04 (0.792)		0.60 (0.831)
Income: High		1.15 (0.886)		0.71 (0.882)		0.16 (0.949)
Education: High		0.23 (0.619)		-1.37** (0.610)		2.21*** (0.823)
Country fixed effects	No	Yes	No	Yes	No	Yes
Own contribution treatment	No	Yes	No	Yes	No	Yes
Root MSE	26.21	26.12	25.59	25.44		
F-test of excluded instrument			7.05	6.90		
Observations	8,498	8,497	8,498	8,497	8,496	8,495

This table reports coefficients and robust standard errors in parentheses (** $p < 0.01$, ** $p < 0.05$, * $p < 0.10$). Columns 1 and 2 report reduced form ordinary least squares (OLS) regression results of *Own Contribution* on the *Other Contribution Treatment: High* which is equal to 1 if the randomized example of the other lottery winner's contribution is equal to 30, 60, or 90 and equal to zero if it is 10. Columns 3 and 4 report the first stage regression of *Expected Contribution* on *Other Contribution Treatment: High*. Columns 5 and 6 report the instrumental variable estimates of *Own Contribution* on *Expected Contribution* using *Other Contribution Treatment: High* to instrument for *Expected Contribution*.

relationship. Individuals with particular unobserved characteristics could tend to be both more likely to think others will contribute generously and therefore, they would contribute more themselves making the reciprocity account of the observed correlation spurious. To estimate the causal effect of expected contribution behavior in representative samples, we exploit the randomized experiment embedded in the instructions for the public goods game. We regress an individual's own contribution decision on the treatment indicator *Other Contribution Treatment: High* that equals 1 if the respondent received instructions in which the other winner's contribution was 30, 60, or 90 €/£/\$ and is zero otherwise. The results are identical when we create indicator variables for all treatment conditions (see Table A5). Model 1 in Table 1 shows the reduced form results. We find that individuals in the high other contribution treatment contribute 2.4 €/£/\$ more than respondents in the low

other contribution treatment. This estimate is almost identical when we add a full set of socio-demographic covariates, country fixed effects, and controls for the own contribution treatment frame.

To estimate the causal effect of expectations about the other winner's contribution on own contributions, we use *Other Contribution Treatment: High* as an instrument for *Expected Contribution*. This instrumental variable (IV) estimation assumes that (a) the randomly assigned treatment *Other Contribution Treatment: High* encourages respondents to have higher expectations about the other winner's contribution and (b) that there is no other direct effect of this treatment on own contributions. Models 3 and 4 in Table 1 report results from a regression of *Expected Contribution* on *Other Contribution Treatment: High* that support the first assumption. We find that individuals in the high other contribution treatment expect about 1.7 €/£/\$ higher contributions from the other winner than respondents in the control group. Although the second assumption (the exclusion restriction) cannot be tested, it appears plausible since the treatment is a randomly assigned specific suggestion about the other winner's contribution.

Models 5 and 6 in Table 1 report the instrumental variable estimates of the causal effect of *Expected Contribution* on *Own Contribution*. A 1 €/£/\$ increase in the contribution that an individual expects from the other winner increases own contributions by about 1.4 €/£/\$. This estimate does not mean that individuals systematically over-reciprocate because when we construct a 95% confidence interval for this estimate, we find that the estimate is not significantly greater than 1 (the interval is (0.65, 2.11)). The effect size is close to what previous lab-experimental work has reported (Fischbacher and Gächter, 2010) and even closer to results based on a heterogeneous sample of adults in Denmark (Thöni et al., 2012).⁷ A natural interpretation of these findings is that cooperative expectations play an important causal role in explaining behavior in social dilemma situations, while socio-demographic characteristics of individuals are not particularly important. At the same time, however, we do not know whether some reciprocity types are more or less widespread among specific socio-demographic groups.

The socio-demographic distribution of reciprocity types

To explore the socio-demographic variation in the use of reciprocal strategies, Figure 4 shows the estimated change in the probability of observing a given strategy type for a given demographic change holding all other variables at their means (Table A7 reports the multinomial logit estimates used for these simulations).

⁷In additional analyses, we also explored whether our other contribution experiment increased the contributions respondents indicated to be willing to make in the strategy method part of our survey. When we estimate the treatment effect, we find that those in the high other contributions treatment increase their own contribution in the strategy part by 2.5 €/£/\$, which is almost exactly the estimate reported above. The IV estimate is 1.45 €/£/\$, which is again almost identical to the effect we estimate when using respondents' initial contribution choice. However, this level effect does not affect the classification of respondents as positive reciprocity types in the results reported below.

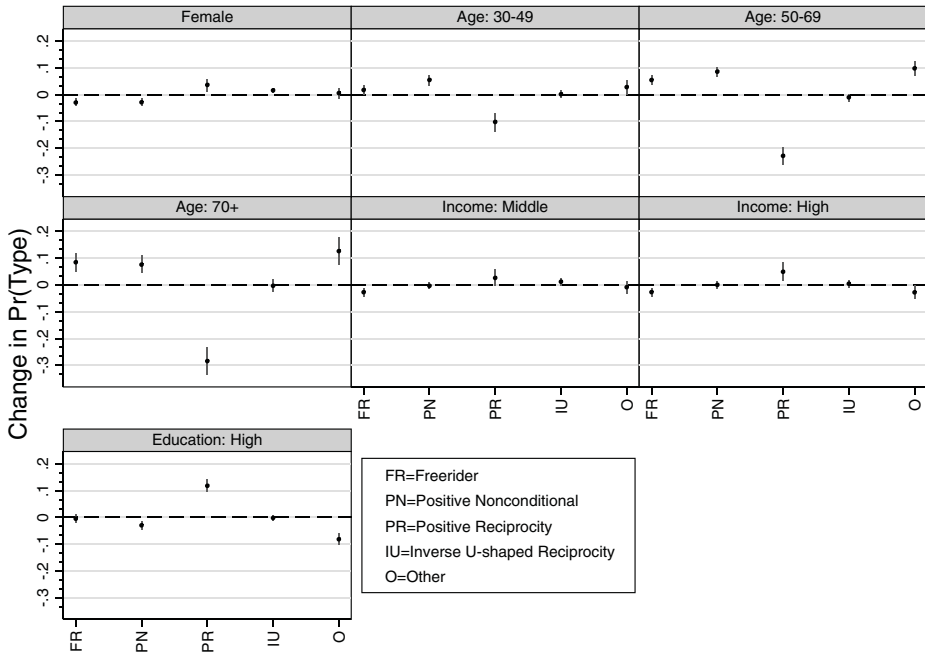


Figure 4

The Socio-demographics of Strategy Types in France, Germany, the United Kingdom, and the United States (Pooled Data).

This plot shows how the probability of a specific strategy type responds to a change in socio-demographic variables together with 95% confidence intervals computed from heteroskedasticity-robust standard errors. The simulations are based on results from a multinomial logistic regression (see Table A7) and were implemented using Clarify (King et al. 2000). For *Female*, the simulated change is from male to female respondents. For all age variables, the reference group is *Age: <30*. For all income variables, the reference groups is *Income: Low*. For *Education: High*, the reference group is *Education: Low*.

We find that female, younger, wealthier, and highly educated respondents are all systematically more likely to use positive reciprocity strategies. For example, the probability of Positive Reciprocity is 11 percentage points higher among high education respondents compared to the lower educated reference group. Overall, these results suggest that a group’s socio-demographic composition matters for which strategy types are more prevalent. Variation in the socio-demographic composition of strategy types leaves open the possibility that some demographic groups may find it easier to cooperate than others. More specifically, groups for which positive reciprocity strategies are more prevalent may be more likely to respond to policy and institutional interventions that seek to raise expectations about the cooperative behavior others.

To explore these socio-demographic differences further, we investigate whether among the plurality of individuals who employ positive reciprocal strategies, there is variation in the responsiveness of cooperative behavior to changes in expectations. Although positive reciprocity generally facilitates cooperative behavior, the long-term evolution of cooperation depends on the exact elasticity of

conditional cooperation, i.e., the extent to which an individual's own contribution to a public good increases in response to a unit increase in expected contributions by others. The higher the elasticity of conditional cooperation, the better the prospects for lasting cooperation (Fischbacher and Gächter, 2010).⁸

To estimate the socio-demographic distribution of contribution elasticity, we first estimated an auxiliary regression for each respondent in which we regressed her/his contribution on a variable that indicated the amount given by the other person (0, 25, 50, 75, and 100 €/£/\$).⁹ We then model an individual's contribution elasticity as a function of a full set of income, age, gender, and education indicator variables. Model 1 in Table 2 shows the results. We find significant differences in the distribution of contribution elasticity across socio-demographic groups. On average, female individuals have lower elasticities than male respondents. We also find, in line with our results on the socio-demographic distribution of strategy types, that older individuals exhibit significantly less strongly reciprocal contribution schedules. Finally, those with higher levels of income and the more educated have significantly larger contribution elasticities. These findings remain virtually identical when adding additional covariates and when re-estimating the results using a tobit model. Overall, even when focusing on those parts of society that generally employ conditionally cooperative strategies, we find that some socio-demographic groups—male, younger, richer, and more educated individuals—appear to be significantly more conditionally cooperative than others.

The causal effect of expectations by strategy type

We now explore whether strategy type conditions the cooperation-enhancing effect of expectations about the contribution behavior of others in theoretically meaningful ways. Specifically, we expect that the causal effect of expected cooperation should be most evident for Positive Reciprocity types. In contrast, the level of cooperation an individual expects should not matter for those employing a freerider strategy. To evaluate these predictions we reestimate the instrumental variable models presented in Table 1 by strategy type. The results reported in Table 3 support our conjecture. The causal effect of expected contribution is strongest among positive reciprocity types with a highly significant IV estimate of 1.3. In contrast, the estimates for Freerider, Inverse U-shaped Reciprocity, and Other are all statistically insignificant. For the remaining category of Positive Nonconditional types, the IV coefficient is smaller (0.89) and only marginally significant.¹⁰

⁸Cooperation failure can result from imperfectly conditional cooperation when for any additional unit contributed by the other, an individual contributes less than one unit. If these actors play against each other repeatedly, contributions will converge to zero over time, a prediction supported by previous laboratory results (Fischbacher and Gächter, 2010).

⁹See Kurzban and Houser (2005) for a comparable approach to determine the contribution elasticity.

¹⁰We also evaluated the importance of strategy type for explaining the effect of expectations about other contributions by reestimating our baseline linear regressions of *Own Contribution* on expected

Table 2
The Socio-demographic Correlates of Contribution Elasticity

	(1) OLS: Basic	(2) OLS: Extended	(3) Tobit
Female	-0.05*** (0.010)	-0.05*** (0.010)	-0.05*** (0.010)
Age: 30–49	-0.02* (0.013)	-0.02 (0.014)	-0.02 (0.014)
Age: 50–69	-0.10*** (0.014)	-0.10*** (0.015)	-0.10*** (0.015)
Age: 70+	-0.12*** (0.027)	-0.11*** (0.029)	-0.11*** (0.029)
Income: Middle	0.01 (0.014)	0.01 (0.015)	0.01 (0.015)
Income: High	0.03** (0.015)	0.03* (0.016)	0.03* (0.016)
Education: High	0.04*** (0.011)	0.04*** (0.011)	0.04*** (0.011)
Altruism: High		-0.02** (0.011)	-0.02** (0.011)
Married		-0.01 (0.014)	-0.01 (0.014)
Separated		-0.04 (0.037)	-0.04 (0.037)
Divorced		-0.02 (0.020)	-0.02 (0.020)
Widowed		-0.06* (0.033)	-0.06* (0.033)
Dompart		-0.02 (0.017)	-0.02 (0.017)
Unemployed		-0.02 (0.019)	-0.02 (0.019)
Ideology		-0.00 (0.002)	-0.00 (0.002)
Germany	0.01 (0.014)	0.00 (0.014)	0.00 (0.014)
United Kingdom	0.05*** (0.014)	0.05*** (0.014)	0.05*** (0.014)
United States	0.03** (0.014)	0.03** (0.014)	0.03** (0.014)
Constant	0.68*** (0.018)	0.70*** (0.022)	0.70*** (0.022)
Observations	3,672	3,672	3,672
R-squared	0.051	0.053	

This table reports coefficients and robust standard errors (in parentheses) from several regressions of contribution elasticity on socio-demographic variables (***) $p < 0.01$, ** $p < 0.05$, * $p < 0.10$. Columns 1 and 2 report ordinary least squares (OLS) regressions. Column 3 reports tobit estimates.

contributions and socio-demographic characteristics presented in Figure 3 by strategy type. Table A3 reports these results that indicate larger coefficient estimates on our expected contribution measures for positive reciprocity types. This lends additional support to the idea that reciprocity type conditions the cooperation-enhancing effects of expectations about the behavior of others.

Table 3
**The Causal Effects of Cooperative Environment on Own Contributions by Strategy Type:
 Instrumental Variable Results**

	Freerider	Positive Nonconditional	Positive Reciprocity	Inverse U-shaped Reciprocity	Other
Expected Contribution (IV)	−1.55 (3.547)	0.89* (0.484)	1.26** (0.503)	0.20 (7.614)	−6.81 (47.802)
Female	−1.44 (2.318)	−1.55 (2.499)	−0.50 (1.256)	−0.77 (18.738)	−22.13 (136.931)
Age: 30–49	0.91 (4.279)	−2.98 (3.104)	0.97 (0.967)	3.85 (24.637)	−16.66 (117.201)
Age: 50–69	−3.61 (4.821)	−0.61 (3.390)	−0.23 (1.237)	1.45 (3.611)	−26.70 (181.515)
Age: 70+	−6.14 (10.221)	−2.63 (3.972)	−1.15 (1.603)	−0.65 (5.123)	−48.11 (317.082)
Income: Middle	2.22 (3.216)	−0.44 (2.176)	1.33 (1.191)	0.34 (36.647)	6.24 (41.657)
Income: High	−1.07 (3.816)	−0.47 (3.179)	0.21 (1.112)	−3.95 (93.314)	12.09 (71.497)
Education: High	−1.16 (1.915)	4.34*** (1.683)	1.75** (0.808)	3.22 (38.018)	−5.76 (44.019)
County fixed effects	Yes	Yes	Yes	Yes	Yes
Own contribution treatment	Yes	Yes	Yes	Yes	Yes
Observations	927	1,130	3,727	409	2,302

This table reports instrumental variable estimates of *Expected Contribution* on *Own Contribution* using *Other Contribution Treatment: High* to instrument for *Expected Contribution*. Robust standard errors are shown in parentheses (*** $p < 0.01$, ** $p < 0.05$, * $p < 0.10$).

DISCUSSION

Societies in which individuals succeed in solving cooperation problems that allow for profitable exchanges have better prospects to grow and develop. The available lab-experimental studies investigating why groups succeed or fail in solving social dilemmas emphasize that individuals cooperate more than predicted by standard economic theory and that the willingness to cooperate is sensitive to expectations about the willingness of others to cooperate—conditional cooperation is an important feature of human behavior (Fehr and Fischbacher, 2002; Fehr and Schmidt, 2006; Ostrom, 2000). These conclusions and many others in this literature are largely based on the behavior of students and other selected populations in a laboratory setting. Although a great deal has been learned from this research, an obvious concern—one shared by many of the contributors to this literature—is that representative populations might behave differently. A number of previous empirical studies have begun to address this issue by studying behavior in social dilemmas in diverse and even relatively large groups of subjects (see, for example, Baldassari and Grossman, 2013; Habyarimana et al., 2009). With just a few exceptions, however, these existing studies are still based on selected and not nationally representative samples.

Extending our knowledge of how individuals behave in social dilemmas to representative samples has the potential to answer several fundamental questions. For example, are some demographic groups more likely to cooperate and if so why? Is the correlation between public goods contribution and expected cooperation by others causal? And do students and the general population differ in how much they react to the behavior of others? We need answers to these and other related questions to determine a baseline level of cooperative behavior among humans and how difficult it will be for different communities to solve social dilemmas. The answers also matter for understanding the potential effectiveness of different policy or institutional interventions to increase cooperation.

In this study, we start to contribute to this agenda by focusing on behavior in a public goods problem and specifically on the role of cooperative environments in explaining cooperation. Our study, based on representative samples in France, Germany, the United Kingdom, and the United States, provides strong evidence that many important insights about behavior in public goods games observed among students and other selected populations are also evident in representative samples. We find that average contributions as a proportion of initial endowment are quite similar to those reported in studies with student subjects. Most importantly, we find that expectations about the contributions of others are strongly correlated with own contributions—reciprocity is central for understanding public goods contributions. Moreover, we implement a new between-subjects design for establishing that this relationship is causal. Our causal estimate based on representative samples further underscores the importance of reciprocity as promoting cooperation in industrialized countries.

We also provide several results that are either in tension with findings in studies on students and other nonrepresentative samples or have not been previously explored. When coding the distribution of conditionally cooperative strategies that individuals employ we find generally similar patterns across our four countries with Positive Reciprocity types constituting a plurality of respondents (generally between 40% and 50%) and with Freeriders making up about 10% of the distribution. Most importantly, we document that the causal effect of expectations about the contributions of others is only clearly evident among respondents coded as positive reciprocity types. These types are unequally distributed in the population with younger, richer, and more educated individuals more likely to employ positive reciprocal strategies, and thus be more likely to respond to treatments that manipulate expectations about others cooperative behavior. This second set of results suggests the potential usefulness of complementing studies of students and other convenience samples with representative samples when studying conditionally cooperative strategies.

A noteworthy implication of our results is that specific socio-demographic groups within societies and the regions or countries in which they concentrate may enjoy an advantage in solving public goods problems. These individuals react positively to institutions that stabilize expectations about the cooperative behavior of others.

Thus, political institutions will be most effective among socio-demographic groups in which positive reciprocity is more widespread. If, for example, individuals do not play conditionally cooperative strategies, institutions that increase expectations about other individuals' contributions will have at best small effects. In contrast, such institutions will have positive effects in populations that indeed use positively reciprocal strategies. Consequently, policymakers and scientists engaging in the design of institutions to facilitate solving cooperation problems, for example, to reduce pollution, tax fraud, or global warming, can improve the effectiveness of policy interventions by taking into account which types of strategies individuals in the target population employ and how these are distributed.

Finally, our study of conditional cooperation in representative samples also promises to shed some light on the long-term evolution of societies in the presence of multiple equilibria with *ex ante* uncertainty about whether a society will realize a more or less cooperative equilibrium in the long-run (Axelrod, 1984). While we have focused on four industrialized countries, subsequent research could begin to map out the distribution of social norms such as conditional cooperation in less developed countries as this may not only improve our ability to understand the dividing lines in political choices on public goods problems but could also explain more fundamental equilibrium selection phenomena that account for differences in countries' economic, social, and political long-term trajectories.

SUPPLEMENTARY MATERIALS

For supplementary material for this article, please visit Cambridge Journals Online: <https://doi.org/10.1017/XPS.2017.16>.

REFERENCES

- Anderson, Jon, Stephen V. Burks, Jeffrey Carpenter, Lorenz Götte, Karsten Maurer, Daniele Nosenzo, Ruth Potter, Kim Rocha and Aldo Rustichini. 2013. "Self-Selection and Variations in the Laboratory Measurement of Other-Regarding Preferences Across Subject Pools: Evidence from One College Student and Two Adult Samples." *Experimental Economics* 16(2): 170–89.
- Ansolabehere, Stephen and Brian F. Schaffner. 2014. "Does Survey Mode Still Matter? Findings From a 2010 Multi-Mode Comparison." *Political Analysis* 22(3): 285–303.
- Ansolabehere, Stephen and Douglas Rivers. 2013. "Cooperative Survey Research." *Annual Review of Political Science* 16(1): 1–23.
- Axelrod, Robert. 1984. *The Evolution of Cooperation*. New York: Basic Books.
- Baldassari, Delia and Guy Grossman. 2013. "The Effect of Group Attachment and Social Position on Prosocial Behavior. Evidence from Lab-in-the-Field Experiments." *PLoS One* 8: 1–9.
- Barrett, Scott. 2003. *Environment and Statecraft: The Strategy of Environmental Treaty-Making*. Oxford: Oxford University Press.

- Bechtel, Michael M., Federica Genovese and Kenneth F. Scheve. 2017. "Interests, Norms and Support for the Provision of Global Public Goods: The Case of Climate Cooperation." *British Journal of Political Science* (forthcoming). DOI: <https://doi.org/10.1017/S0007123417000205>
- Bechtel, Michael M. and Kenneth F. Scheve. 2013. "Mass Support for Climate Cooperation Depends on Institutional Design." *Proceedings of the National Academy of Sciences of the United States of America* 110(34): 13763–13768.
- Bechtel, Michael M. and Kenneth F. Scheve. 2017. "Replication Data for: 'Who Cooperates? Reciprocity and the Causal Effect of Expected Cooperation in Representative Samples'." *Harvard Dataverse, V3*.
- Bellemare, Charles and Sabine Kröger. 2007. "On Representative Social Capital." *European Economic Review* 51(1): 183–202.
- Belot, Michele, Raymond Duch and Luis Miller. 2015. "A Comprehensive Comparison of Students and Non-students in Classic Experimental Games." *Journal of Economic Behavior and Organization* 113: 26–33.
- Berinsky, Adam J., Gregory A. Huber and Gabriel S. Lenz. 2012. "Evaluating Online Labor Markets for Experimental Research: Amazon.com's Mechanical Turk." *Political Analysis* 20: 351–368.
- Bolsen, Toby, Pail J. Ferraro and Joan Jose Miranda. 2014. "Are Voters More Likely to Contribute to Other Public Goods? Evidence from a Large-Scale Randomized Policy Experiment." *American Journal of Political Science* 58(1): 17–30.
- Camerer, Colin F. and Robin M. Hogarth. 1999. "The Effects of Financial Incentives in Experiments: A Review and Capital-Labor-Production Framework." *Journal of Risk and Uncertainty* 19(1): 7–42.
- Carlsson, Frerik, Olof Johansson-Stenmann and Pham Khanh Nam. 2014. "Social Preferences are Stable Over Long Periods of Time." *Journal of Public Economics* 117: 104–114.
- Charness, Gary and Ernan Haruvy. 2002. "Altruism, Equity, and Reciprocity in a Gift-exchange Experiment: An Encompassing Approach." *Games and Economic Behavior* 40(2): 203–231.
- Crosen, Rachel and Uri Gneezy. 2009. "Gender Differences in Preferences." *Journal of Economic Literature* 47(2): 448–474.
- Dickson, Eric S. 2011. "Economics versus Psychology Experiments: Stylization, Incentives, and Deception." In *Cambridge Handbook of Experimental Political Science*, eds. Donald Green and James N. Druckman. Cambridge: Cambridge University Press, pp. 58–70.
- Druckman, James N. and Cindy D. Kam. 2011. *Students as Experimental Participants: A Defense of the "Narrow Data Base"*. Cambridge: Cambridge University Press, pp. 41–57.
- Engelmann, Dirk and Martin Strobel. 2010. "Inequality Aversion and Reciprocity in Moonlighting Games." *Games* 1(4): 459–477.
- Fehr, Ernst and Klaus M. Schmidt. 2006. "The Economics of Fairness, Reciprocity and Altruism - Experimental Evidence and New Theories." In *Handbook of the Economics of Giving, Altruism and Reciprocity, Volume I*. New York: Elsevier B.V., pp. 615–91.
- Fehr, Ernst and Urs Fischbacher. 2002. "Why Social Preferences Matter - The Impact of Non-Selfish Motives on Competition, Cooperation and Incentives." *Economic Journal* 112(478): C1–C33.
- Fehr, Ernst, Urs Fischbacher, Bernhard von Rosenblatt, Jürgen Schupp and Gert G. Wagner. 2002. "A Nation-Wide Laboratory: Examining Trust and Trustworthiness by Integrating Behavioral Experiments into Representative Surveys." *Schmollers Jahrbuch* 122(4): 519–542.

- Fischbacher, Urs and Simon Gächter. 2010. "Social Preferences, Beliefs, and the Dynamics of Free Riding in Public Goods Experiments." *American Economic Review* 100(1): 541–556.
- Fischbacher, Urs, Simon Gächter and Ernst Fehr. 2001. "Are People Conditionally Cooperative? Evidence from a Public Goods Experiment." *Economics Letters* 71(3): 397–404.
- Frey, Bruno S. and Benno Torgler. 2007. "Tax Morale and Conditional Cooperation." *Journal of Comparative Economics* 35(1): 136–159.
- Gächter, Simon, Benedikt Herrmann and Christian Thöni. 2004. "Trust, Voluntary Cooperation, and Socio-economic Background: Survey and Experimental Evidence." *Journal of Economic Behavior and Organization* 55(4): 505–531.
- Gillis, Mark and Paul Hettler. 2007. "Hypothetical and Real Incentives in the Ultimatum Game and Andreoni's Public Goods Game: An Experimental Study." *Eastern Economic Journal* 33(4): 491–510.
- Habyarimana, James, Macartan Humphreys, Daniel N. Posner and Jeremy M. Weinstein. 2007. "Why Does Ethnic Diversity Undermine Public Goods Provision." *American Political Science Review* 101(4): 709–725.
- Habyarimana, James, Macartan Humphreys, Daniel N. Posner and Jeremy M. Weinstein. 2009. *Coethnicity: Diversity and the Dilemmas of Collective Action*. New York: Russell Sage Foundation.
- Henrich, Joseph, Robert Boyd, Samuel Bowles, Colin Camerer, Ernst Fehr, Herbert Gintis and Richard McElreath. 2001. "In Search of Homo Economicus: Behavioral Experiments in 15 Small-Scale Societies." *American Economic Review* 91(2): 73–78.
- Hollyer, James R., B. Peter Rosendorff and James Raymond Vreeland. 2015. "Transparency, Protest, and Autocratic Instability." *American Political Science Review* 109(4): 764–784.
- Holm, Håkan J. and Anders Danielson. 2005. "Tropic Trust Versus Nordic Trust: Experimental Evidence From Tanzania And Sweden." *Economic Journal* 115(503): 505–532.
- Huff, Connor and Dustin Tingley. 2015. "Who Are These People?" Evaluating the Demographic Characteristics and Political Preferences of MTurk Survey Respondents." *Research and Politics* (1-12).
- Hume, David. 2003 [1740]. *A Treatise of Human Nature*. Oxford: Oxford University Press.
- Kaul, Inge, Isabelle Grunberg and Marc A. Stern. 1999. "Defining Global Public Goods." In *Global Public Goods: International Cooperation in the 21st Century*, eds. Inge Kaul, Isabelle Grunberg and Marc A. Stern. Oxford: Oxford University Press, pp. 2–20.
- Keohane, Robert O. and Elinor Ostrom. 1995. *Local Commons and Global Interdependence. Heterogeneity and Cooperation in Two Domains*. London: Sage Publications.
- Kern, Holger L., Elizabeth A. Stuart, Jennifer Hill and Donald P. Green. 2016. "Assessing Methods for Generalizing Experimental Impact Estimates to Target Populations." *Journal of Research on Educational Effectiveness*, forthcoming.
- King, Gary, Michael Tomz and Jason Wittenberg. 2000. "Making the Most of Statistical Analysis: Improving Interpretation and Presentation." *American Journal of Political Science* 44(2): 347–361.
- Kurzban, Robert and Daniel Houser. 2005. "Experiments Investigating Cooperative Types in Humans: A Complement to Evolutionary Theory and Simulations." *Proceedings of the National Academy of Sciences* 102(5): 1803–1807.
- Morton, Rebecca B. and Kenneth C. Williams. 2010. *Experimental Political Science and the Study of Causality: From Nature to the Lab*. Cambridge, MA: Cambridge University Press.

- Olson, Mancur. 1965. *The Logic of Collective Action: Public Goods and the Theory of Groups*. Cambridge, MA: Harvard University Press.
- Ostrom, E. 2000. "Collective Action and the Evolution of Social Norms." *Journal of Economic Perspectives* 14: 137–158.
- Rauhut, Heiko and Fabian Winter. 2010. "A Sociological Perspective on Measuring Social Norms by Means of Strategy Method Experiments." *Social Science Research* 31(6): 1181–1194.
- Rivers, Douglas. 2011. "Sample Matching: Representative Sampling from Internet Panels." YouGov White Paper.
- Roux, Catherine and Christian Thöni. 2015. "Do Control Questions Influence Behavior in Experiments?" *Experimental Economics* 18: 185–194.
- Sandler, Todd and Keith Hartley. 1999. *The Political Economy of NATO: Past, Present and into the 21st Century*. Cambridge: Cambridge University Press.
- Selten, Reinhard. 1967. "Die Strategiemethode zur Erforschung des eingeschränkt rationalen Verhaltens im Rahmen eines Oligopol-experiments." In *Beiträge zur experimentellen Wirtschaftsforschung*, ed. H. Sauermann. Tübingen: J.C.B. Mohr (Paul Siebeck), pp. 136–168.
- Smith, Richard, Robert Beaglehole, David Woodward and Nick Drager, eds., 2003. *Global Public Goods for Health: A Health Economic and Public Health Perspective*. Oxford: Oxford University Press.
- Stiglitz, Joseph E. 1999. "Knowledge as a Global Public Good." In *Global Public Goods: International Cooperation in the 21st Century*, eds. Inge Kaul, Isabelle Grunberg and Marc Stern. Oxford: Oxford University Press, pp. 308–325.
- Stoop, Jan, Charles N. Noussair and Daan van Soest. 2012. "From the Lab to the Field: Cooperation among Fishermen." *Journal of Political Economy* 120(6): 1027–1056.
- Thöni, Christian, Jean-Robert Tyran and Erik Wengström. 2012. "Microfoundations of Social Capital." *Journal of Public Economics* 96(7-8): 635–643.
- Tingley, Dustin and Michael Tomz. 2014. "Conditional Cooperation and Climate Change." *Comparative Political Studies* 47(3): 344–368.
- Tsai, Lily L. 2007. "Solidary Groups, Informal Accountability, and Local Public Goods Provision in Rural China." *American Political Science Review* 101(2): 355–372.
- Ward, Hugh. 1990. "Three Men in a Boat, Two Must Row: An Analysis of a Three-Person Chicken Pregame." *Journal of Conflict Resolution* 34(3): 371–400.
- West, Stephen G., Naihua Duan, Willo Pequegnat, Paul Gaist, Don C. Des Jarlais, David Holtgrave, José Szapocznik, Martin Fishbein, Bruce Rapkin, Michael Clatts and Patricia Dolan Mullen. 2008. "Alternatives to the Randomized Controlled Trial." *American Journal of Public Health* 98(8): 1359–1366.