# Do Lab Experiments Misrepresent Social Preferences? The Case of Self-Selected Student Samples* 

Armin Falk ${ }^{\dagger}$ Stephan Meier $\ddagger$ and Christian Zehnder ${ }^{\S}$

October 10, 2011


#### Abstract

Social preference research has received considerable attention among economists in recent years. However, the empirical foundation of social preferences is largely based on laboratory experiments with self-selected students as participants. This is potentially problematic as students participating in experiments may behave systematically different than non-participating students or non-students. In this paper we empirically investigate whether laboratory experiments with student samples misrepresent the importance of social preferences. Our first study shows that students who exhibit stronger prosocial inclinations in an unrelated field donation are not more likely to participate in experiments. This suggests that self-selection of more prosocial students into experiments is not a major issue. Our second study compares behavior of students and participants recruited from the general population in a trust experiment. In general, we find very similar behavioral patterns for the two groups, but nonstudents make significantly more generous repayments suggesting that results from student samples might be seen as a lower bound for the importance of prosocial behavior.


Keywords: methodology, selection, experiments, prosocial behavior
JEL: C90, D03

[^0]
## 1. Introduction

Social preferences such as concerns for distributional fairness and reciprocity have received considerable attention in recent economic research (see, e.g., Cooper and Kagel, forthcoming). The empirical foundation of social preferences is largely based on laboratory experiments using self-selected students as samples. This is a potential problem, as students participating in experiments might behave systematically different than non-participating students or nonstudents. If participating students behave more or less prosocially than the population of interest, our laboratory results provide a biased estimation of the potential of social preferences for the analysis of economic outcomes. Were this to be the case we would need to be more careful in plugging behavioral assumptions derived from observations in the lab into models used to derive implications for the general population.

In this paper we provide empirical evidence on whether laboratory experiments with student samples systematically misrepresent social preferences. In particular, we address two potential problems: First, experiments rely on volunteers, creating a problem of self-selection. This may bias outcomes in experiments if participants exhibit a stronger or weaker prosocial inclination than people who do not participate. A priori, the direction of a potential selection effect is unclear. If people's participation decision is mainly money driven, one might expect an overrepresentation of self-interested payoff-maximizers in the participant pool. However, it is also possible that social motives determine people's decision to participate (e.g., people may want to help the researcher or foster the advancement of science), which would
speak for an overrepresentation of prosocially inclined participants. ${ }^{1}$ While a drastic overestimation of prosocial motives would be especially troubling for the literature on social preferences, it is, of course, also important to know whether there is a bias in the other direction. Second, most laboratory experiments are conducted with university undergraduates. While using students as subjects is very convenient, they are not representative of the general population in many dimensions. The important question for our context is whether they also differ with respect to social preferences, so that using them as participants distorts the measurements of social preferences in experiments.

Our first study analyzes whether participating students are more prosocial than non-participating students. The ideal data set to test for potential differences between participants and non-participants would provide information on prosocial preferences of all students while observing who participates in experiments and who does not. This type of data is usually not available simply because we have proxies for preferences typically only for participants in experiments. Moreover, if we know preferences from nonexperimental data, e.g., survey studies, we do not observe decisions to participate in an experiment. In our first study we present results using a novel data set that combines preference measures for both participants and non-participants. In particular, we use a naturally occurring donation decision as a measure of participants' and non-participants' prosocial inclination. Our results show that students with stronger prosocial inclinations are neither

[^1]more likely to participate in experiments (extensive margin), nor do they participate more often (intensive margin). These findings resonate with a complementary study by Cleave et al. (2010) who also don't find a selectionbias regarding social preferences. However, while Cleave et al. (2010) make use of tutorials of introductory microeconomics to obtain a laboratory measure of social preferences for about 600 students, we identify prosocial inclinations using a naturally occurring field donation that gives us access to data for more than $16^{\prime} 000$ students. While both approaches have their advantages, the fact that both studies ultimately emphasize a non-result makes a large number of observations relevant, because it increases the precision of the estimation and reduces the possibility to find a null-effect by chance. In fact, we show that our sample allows us to estimate the null result with a small confidence interval.

Our second study uses a version of the trust game (Berg et al., 1995) to investigate whether measurements of social preferences change if the usual student subjects are replaced with participants from the general population. In contrast to many existing studies, we use the same recruitment procedure, the same instructions, the same decision process and the same financial incentives for both our subject pools. Our results reveal no significant difference in first mover trusting behavior between students and non-students. However, the repayment level is significantly lower for students than for non-students. Our results are in line with earlier studies that also show that prosocial behavior is even more frequently observed with non-student participants (see e.g., Fehr and List, 2004; Bellemare and Kröger, 2007; Dohmen et al., 2008; Burks et al., 2009; Belot et al., 2010).

Our paper contributes to a recent methodological debate about the role of experimental economics in the social sciences (see, e.g., Levitt and List, 2007; Falk and Heckman, 2009; List, 2009; Croson and Gächter, 2010; Bardsley et al., 2010; Henrich et al., 2010; Gächter, 2010). Some of this work has raised serious concerns about the relevance of lab findings with regard to the role of social preferences. This paper provides a step in empirically investigating one issue raised in this debate. Our results suggest that using self-selected student samples does not contribute to a systematic overestimation of social preferences. On the contrary, the results of our second study indicates that results obtained from student samples might be seen as a lower bound for the importance of prosocial behavior. Of course, our results do not exclude that laboratory experiments may provide distorted estimates of social preferences for other reasons (such as low stakes, short durations, high degrees of scrutiny). However, we see our paper as a starting point and hope that future research will investigate the empirical relevance of other potential sources for biases.

The paper is organized as follows. Section 2 contains the field study on selection of students into experiments. The question of whether students and non-students have different prosocial inclinations is discussed in section 3. Section 4 concludes.

## 2. Do Social Preferences Predict Self-Selection?

### 2.1. Research Design

This section analyzes whether self-selection of students into experiments leads to a misrepresentation of prosocial preferences in the participating part
of the student population. We study decisions of students to participate in experiments organized by the experimental economics laboratory of the University of Zurich. Our sample consists of 16,666 undergraduates who registered at the University of Zurich between the fall term 1998 and the spring term 2004 and for whom registration at the University of Zurich is their first enrollment at a University. For all those students, we know whether and how often they participated in an economics experiment between the fall term 1998 and the fall term 2005. In total 1,783 students participated at least once, i.e., the participation rate is about 11 percent. Conditional on participating at least once, the students participate in 2.5 experiments on average.

To measure the extent of all students' prosocial inclinations we use a naturally occurring prosocial decision at the University of Zurich as a proxy. Each semester, every student has to decide whether or not he or she wants to contribute a pre-determined amount to two social funds which provide charitable services (financial support for foreign students (CHF 5) and free loans for needy students (CHF 7), for further details, see Frey and Meier, 2004a,b, CHF $1 \sim$ USD 0.85). Students can therefore give CHF $0,5,7$ or 12 (both funds together). The level of possible donations is thus very similar to stake sizes typically used in lab settings.

There are several features why these donation decisions constitute an interesting proxy for social preferences. First, the measure does not rely on selfreported survey responses but on actual decisions. Second, donation decisions
are made in private and never made public. ${ }^{2}$ Third, students are unaware that their behavior is analyzed in a research study. Fourth, and most importantly, all students at the university have to decide about the donations. Thus, our measure is not subject to any selection issue.

However, as with most field measures there are also potential problems. Since the persuasive power of our results critically depends on the quality of our measurement of social preferences, it is important to discuss in detail the different measures we use and how they address potential caveats. Our first measure (First Field Donation) only considers a student's donation decision when he or she first registers for a program. This measure has several advantages. First, the university rules require that each student has to show up in person at the registration office for the initial enrollment. This ensures that we know with certainty that this first donation decision has been made by the student him- or herself. Second, as the initial enrollment takes place before the first semester starts, this measure is collected before students have taken any courses at the University, before they have been exposed to any lab recruitment efforts and before they have participated in any experiment. We can therefore rule out the possibility of reversed causality as participation in experiments cannot have influenced the decision to contribute to the funds. These features make this measurement a particularly clean one.

[^2]Our second measure (Average Field Donation) exploits information on all donation decisions taken by a student. For each individual, we calculate the average donation amount over all observed contributions. Using several measures per individual has the advantage of reduced measurement error. A potential problem is that the forms for registration renewals can be completed at home. Therefore, we cannot be sure that it is the student him- or herself who fills out the form. However, because students also have to provide details regarding major and minor study subjects on the same form, it is quite unlikely that another person can perform this task. To further increase the confidence that the variable Average Field Donation measures an individual's prosocial inclination we use data collected by Benz and Meier (2008). They perform a modified dictator game in the laboratory using a subsample of the students in our data set as participants. It turns out that individuals with higher average field donations transfer a significantly higher share of their endowment to the recipient (Spearman's Rho $=0.29, p<0.0001$ ). This provides direct evidence that our field measure captures the same social motivations as the simple experiments typically used in the laboratory. Finally, it is also reassuring to notice that our two measures First Field Donation and Average Field Donation are strongly correlated (Spearman's Rho $=0.73, p<0.001$ ). ${ }^{3}$

[^3]
### 2.2. Results

Panel A in Table 1 reveals that participants differ in various dimensions from non-participants. These differences indicate the relevance of self-selection of particular groups of students. In Panel B of Table 1 we investigate whether this selection is also associated with differences in prosocial inclinations. The panel provides descriptive statistics of contributions to the two funds for participants and non-participants. The summary statistic does not show any significant difference between participants and non-participants. In their first decision, the same proportion of participants and of non-participants contributed to at least one of the two funds ( 75 percent) and, on average, they donate about the same amount (CHF 8.39 vs. $8.45 ; p=0.67$ in a $t$-test). Figure 1 illustrates that both the participation rate and the number of experiments a student participated in does not significantly depend on individuals' first donation decisions. None of the differences are statistically significant. When we look at all decisions of a student, it turns out that participants contribute on average in 77 percent of all decisions, while non-participants' contribution rate is 76 percent (n.s.). There is also no substantial difference in the average amount donated (CHF 8.66 vs. 8.84; $p=0.09$ in a $t$-test; see also the distribution of average donations in Web Appendix Figure A1). Thus, the raw data analysis does not reveal any significant difference in prosocial inclinations of participants and non-participants.

Panel A of Table 2 reports Probit estimations, where the dependent variable is an indicator variable for the decision to participate in experiments and the independent variable is either the first donation (columns (1), (2), and (3))
or the average field donation (columns (4), (5), and (6)). ${ }^{4}$ We report marginal effects in brackets. Column (1) shows that students who contribute more money in their first decision are not significantly more likely to participate in an experiment than those who don't. The marginal effect is essentially zero. As a consequence of the large number of observations, our effects are quite precisely estimated. The $95 \%$ confidence interval of the marginal effect is $[-0.1,0.1]$ (in percentage points). This implies that a change in the magnitude of one standard deviation in the first donation decision (s.d. $=5.2$ ) is very unlikely to increase (decrease) the participation rate by more than 0.6 (0.4) percentage points (i.e., an increase (decrease) of 5.6 (3.7) percent relative to the average participation rate of $10.7 \%$ ).

Column (4) reports a regression using the Average Field Donation as a proxy for prosocial inclinations. This proxy is potentially influenced by students' experience at the University including their participation in experimental studies. The results are very similar to the ones obtained from using only the first decision: Individuals who contribute on average more to the charitable funds are not significantly more likely to participate in experiments. The marginal effects indicate that the participation rate of students who contribute on average one CHF more is only about 0.1 percentage points higher. This means that for an increase in the average field donation of one standard deviation $($ s.d. $=4.1)$, the participation rate increases by only 0.4 percentage points (i.e., an increase of 3.7 percent relative to the average participation rate of $10.7 \%$ ). Given the large number of observations the lack of a significant effect

[^4]is a strong result. The $95 \%$ confidence interval of the marginal effect is $[-0.02$, 0.2 ] (in percentage points) indicating that it is extremely unlikely that changing the average field donation by one s.d. increases (decreases) the participation rates by more than $0.9(0.08)$ percentage points. ${ }^{5}$

In addition to participating for a first time, it is also interesting to investigate if social preferences predict whether a student becomes a regular participant. ${ }^{6}$ Column (7) and (8) show Tobit regressions with the number of experiments an individual participated in as dependent variable. The estimations show that both the 'First Field Donation' and the 'Average Field Donation' are not good predictors for how often somebody participates in experiments (this holds both overall and conditional on participating, see Web Appendix Table A3 for additional specifications).

As the main purpose of this study is to detect differences between populations (and not to explain these differences if they exist), the estimations without controls are the most important ones. The descriptive statistics reveal many significant differences between the two groups of interest (e.g., gender and major). The question that we want to answer is: do these differences also imply that there is a difference regarding social preferences between these groups? To answer this question, it is important not to include controls (because the

[^5]observable heterogeneity may exactly be the reason for the difference in social preferences). Therefore, Columns (1), (4), (7) and (8) contain our main results.

However, it can be of separate interest whether there is selection for certain groups. To investigate this question, we add two types of controls. In column (2) and (5) we add 'demographic' variables (gender, age, foreigner status, number of semesters, cohort dummies). The results don't change. In columns (3) and (6) we additionally control for the field of study. While the marginal effect doesn't change it becomes significant at the $5 \%$-level. ${ }^{7}$ This indicates that for certain majors, participants may select based on their field donation. Panel B of Table 2 shows separate regressions for different subgroups that might be interesting for research on prosocial behavior. The results show that the marginal effect is bigger for men than for women, but none is significant. The effects also remains insignificant if we consider economists and non-economists separately. If we estimate the effect for the field of studies that are most represented in experiments (law and arts), we find a significant effect for students from the arts faculty.

In sum, our results do not support the hypothesis that participating students have different social preferences than non-participants. This suggests that within the group of students the bias due to self-selection on social preferences is likely to be small. While there might be some selection within certain subgroups, these subgroups do not make up a sufficient part of a typical

[^6]student sample to yield an overall significant effect. However, it is still possible that student participants behave differently than participants recruited from a more general subject pool. We investigate this question in the next section.

## 3. Do Students Have Different Social Preferences?

### 3.1. Research Design

We conduct two identical trust experiments using distinct subject pools for the recruitment of participants. Contrary to most existing studies, we use the same recruitment procedure, the same instructions, the same decision process and the same financial incentives for participants in both experiments. Therefore differences in prosocial behavior can only be caused by differences between the two subject pools. All participants in the experiments live in Zurich. However, while one group of our participants was recruited from the student pool at the University of Zurich, the other group was recruited from a representative sample of the population of the city of Zurich (for details on the recruitment procedure of this study, see Web Appendix).

As participation was voluntary, both our groups of participants are selfselected. In light of our first study it seems plausible to assume the absence of important selection effects with respect to social preferences, but we cannot directly rule out such a possibility with our data. However, our results are informative in any case. Even if sorting takes place our study tells us whether recruiting subjects from the general population yields a different measurement of prosocial inclinations than recruiting subjects from a student pool. This is
of practical importance as the vast majority of experiments and surveys relies on voluntary participation.

To measure social preferences, we use a variant of the trust game (Berg et al., 1995). Both subjects receive an endowment of CHF 20. The first mover decides how much of his endowment to transfer to the second mover. The transfer can be any amount in steps of 2 CHF , i.e., $0,2,4, \ldots$, or 20 CHF. The chosen transfer is tripled by the experimenter and passed to the second mover. Contingent upon the first mover's transfer the second mover decides on a back transfer. This back transfer can be any integer amount between 0 and 80 CHF. The first mover earns his endowment minus his own transfer plus the back transfer of the second mover. The second mover gets his endowment plus three times the first mover's transfer minus the back transfer. ${ }^{8}$

In order to elicit second movers' willingness to reciprocate, we used the contingent response method (see Brandts and Charness, 2011, for a discussion about the validity of the method). This means that each second mover, before knowing the actual first mover's investment, made a back transfer decision for each of the 11 possible investments $(0,2, \ldots, 20)$ of the first mover. The advantage of the contingent response method is that it allows us to measure each second mover's willingness to reciprocate independently of the transfer which he actually received. This is important, because it enables us to make a clean comparison of the level of reciprocity, even if first movers

[^7]behave differently between subject pools (for details on the procedure, see Web Appendix).

### 3.2. Results

In total we have 1296 participants in the experiment (295 recruited from the student pool, 1001 recruited from the general population). Students and non-students differ in many socio-demographic dimensions. In particular, we observe that non-students are on average older, more likely to be married, less well educated, and more likely to be right-wingers (see Table A4 in the Web Appendix). In this study we investigate whether students and non-students also exhibit different prosocial inclinations. We start by examining trusting behavior of first movers. A simple comparison of first mover transfers between the two groups reveals only a small difference across the two subject pools (13.17 for non-students vs. 13.47 for students). An OLS regression of first mover transfers on a student dummy (column (1) of Table 3) reveals that the observed difference of 0.30 is not statistically significant. ${ }^{9}$ The $95 \%$ confidence interval for this effect is $[-0.9,1.5]$. This reveals that it is very unlikely that first mover transfers of the two groups differ by more than about $10 \%$. While the uncontrolled regression is the most relevant for our comparison of subject pools, it is also of interest to investigate the role of observable differences. Including control variables allows us to compare participants from the student pool to participants from the general populations with similar socio-demographic backgrounds. Adding control variables changes the sign of the student coefficient, but the effect

[^8]remains insignificant (see column (2)). ${ }^{10}$ Results in column (3) and (4) show that the decisions of students and non-students are not driven by different beliefs about the behavior of second movers.

We now turn to second movers' behavior. Figure 2 shows the average second mover back transfers conditional on first mover transfer. For every possible first mover transfer students make lower average repayments than non-students. All differences are statistically significant (see Web Appendix Table A6 for the corresponding $p$-values). Averaging over all backtransfers, students transfer back 15 percent less than non-students. The fact that students transfer back less than non-students does not imply that they generally react less sensitive to first movers' transfers. In fact Figure 2 illustrates that the slope between first mover transfer and second mover back transfer is very similar. Put differently, students' and non-students' reciprocation pattern is very similar; the only difference being that students reciprocate on a lower absolute level. Column (5) of Table 3 confirms this. It shows an OLS regression with second movers' back transfers as the dependent variable. We regress back transfers on a student dummy, the first mover transfer and the interaction effect between student dummy and first mover transfer. The coefficient of the student dummy is negative and significant, i.e., students transfer back significantly less than non-students. However, the interaction effect is close to zero indicating that

[^9]students and non-students exhibit a similar reciprocal inclination as suggested by Figure 2. If we add socio-demographic controls to the regression (see column (6)), the coefficient of the student dummy is no longer significant. This indicates that students are not less prosocial than other participants with a similar sociodemographic background, i.e., the difference between the subject pools is driven by the fact that students and non-students differ with regard to their sociodemographic background. ${ }^{11}$

## 4. Concluding remarks

This paper empirically tests whether laboratory experiments with students systematically misrepresent the importance of social preferences. Such an empirical test is critical as experimental methods become increasingly important in economics and experimental results, especially those on social preferences, often challenge insights and policy implications of standard economic models.

Our first study shows that the degree of prosocial behavior in an unrelated field donation does not predict whether (and how often) students participate in experiments. This suggests that self-selection does not significantly bias the social preferences measured in the laboratory. The results of our second study reveal that student participants and non-student subjects show very similar behavioral patterns in our trust experiment. While students make less

[^10]generous repayments, their investment behavior, their beliefs about second mover behavior, and their reciprocal inclination are very similar to those of participants recruited from the general population.

## References

Bardsley, Nicholas, Robin Cubitt, Graham Loomes, Peter Moffatt, Chris Starmer, and Robert Sugden (2010). Experimental economics: Rethinking the rules. Princeton University Press.
Bellemare, Charles and Sabine Kröger (2007). "On representative social capital." European Economic Review, 51(1), 183-202.
Belot, Michèle, Raymond Duch, and Luis Miller (2010). "Who should be called to the lab? A comprehensive comparison of students and non-students in classic experimental games." Discussion Papers 2010001, University of Oxford, Nuffield College, URL http://ideas.repec.org/p/cex/dpaper/2010001.html.
Benz, Matthias and Stephan Meier (2008). "Do People Behave in Experiments as in the Field? Evidence from Donations." Experimental Economics, 11(3), 268-281.
Berg, Joyce, John Dickhaut, and Kevin McCabe (1995). "Trust, Reciprocity, and Social History." Games and Economic Behavior, 10(1), 122-142.
Brandts, Jordi and Gary Charness (2011). "The strategy versus the direct-response method: a first survey of experimental comparisons." Experimental Economics, p. Forthcoming.
Burks, Stephen, Jeffrey Carpenter, and Lorenz Götte (2009). "Performance pay and worker cooperation: Evidence from an artefactual field experiment." Journal of Economic Behavior $\mathcal{E}^{3}$ Organization, 70(3), 458-469.
Cleave, Blair L., Nikos Nikiforakis, and Robert Slonim (2010). "Is There Selection Bias in Laboratory Experiments?" Department of Economics - Working Papers Series 1106, The University of Melbourne, URL http://econpapers.repec.org/ RePEc:mlb:wpaper:1106.
Cooper, David and John Kagel (forthcoming). The Handbook of Experimental Economics, Vol. 2, chap. Other Regarding Preferences: A Selective Survey of Experimental Results.
Croson, Rachel and Simon Gächter (2010). "The science of experimental economics." Journal of Economic Behavior \& Organization, 73(1), 122-131.
Dohmen, Thomas, Armin Falk, David Huffman, and Uwe Sunde (2008). "Representative Trust And Reciprocity: Prevalence And Determinants." Economic Inquiry, 46(1), 84-90.
Falk, Armin and James Heckman (2009). "Lab Experiments Are a Major Source of Knowledge in the Social Sciences." Science, 326(5952), 535-538.
Fehr, Ernst and John A. List (2004). "The Hidden Costs and Returns of Incentives - Trust and Trustworthiness among CEOs." Journal of the European Economic Association, 2(5), 743-71.
Frey, Bruno S. and Stephan Meier (2004a). "Pro-Social Behavior in a Natural Setting." Journal of Economic Behavior and Organization, 54(1), 65-88.
Frey, Bruno S. and Stephan Meier (2004b). "Social Comparison and Pro-Social Behavior: Testing Conditional Cooperation in a Field Experiment." American Economic Review, 94, 1717-22.

Gächter, S. (2010). "(Dis)advantages of student subjects: What is your research question?" Behavioral and Brain Sciences, 33(2-3), 92-93.
Henrich, J., S.J. Heine, and A. Norenzayan (2010). "The weirdest people in the world." Behavioral and Brain Sciences, 33(2-3), 61-83.
Levitt, Steven D. and John A. List (2007). "What Do Laboratory Experiments Measuring Social Preferences Reveal About the Real World?" Journal of Economic Perspectives, 21(2), 153-174.
List, John A. (2009). "Social Preferences: Some Thoughts from the Field." Annual Review of Economics, 1(1), 1:21.1-21.20.

Figure 1. First Field Donation and Participation in Experiments


Note: The figure shows the participation rate (left axis) and the average number of experiments a student participated in (including students who did never participate, right axis) depending on the first field donation in study 1. Distribution of First Field Donation: $25.20 \%$ contribute CHF $0,4.19 \%$ contribute CHF $5,5.68 \%$ contribute CHF 7 , and $64.93 \%$ contribute CHF 12.

Figure 2. Back transfers of Students and Non-Students in Field Trust Game


Note: The figure shows average repayments of second movers in the trust game of study 2 .
The lower line depicts average repayments of participants recruited from the student subject pool of the University of Zurich. The upper line depicts average repayments of participants recruited from a representative sample of the average population of the city of Zurich.

Table 1. Summary Statistics of Study 1

|  | Non-participants |  | Participants |  | $t$-test $/$ |
| :--- | :---: | :---: | :---: | :---: | :---: |
| Variable | Mean | s.d. | Mean | s.d. | $\chi^{2}$-test $^{1}$ |
|  |  |  |  |  |  |
| Panel A: Observable characteristics |  |  |  |  |  |
| Age at registration | 21.94 | 4.21 | 21.07 | 2.87 | $p<0.01$ |
| No. of semesters | 5.34 | 3.26 | 5.97 | 3.15 | $p<0.01$ |
| Gender (Women=1) | 0.57 | 0.50 | 0.53 | 0.50 | $p<0.01$ |
| Nationality (Foreigner=1) | 0.08 | 0.27 | 0.07 | 0.25 | $p<0.05$ |
| Computer science | 0.04 | 0.18 | 0.03 | 0.16 | $p=0.21$ |
| Economics \& Business | 0.13 | 0.32 | 0.14 | 0.34 | $p<0.05$ |
| Theology | 0.01 | 0.08 | 0.003 | 0.05 | $p<0.05$ |
| Law | 0.16 | 0.36 | 0.25 | 0.42 | $p<0.01$ |
| Medicine | 0.07 | 0.26 | 0.18 | 0.38 | $p<0.01$ |
| Veterinary medicine | 0.03 | 0.16 | 0.03 | 0.16 | $p=0.64$ |
| Arts faculty | 0.47 | 0.49 | 0.33 | 0.46 | $p<0.01$ |
| Natural science | 0.10 | 0.30 | 0.05 | 0.21 | $p<0.01$ |
|  |  |  |  |  |  |
| Panel B: Prosocial behavior |  |  |  |  | $p=0.80$ |
| Contributed in first decision $(=1)$ | 0.75 | 0.43 | 0.75 | 0.43 | $p=0.67$ |
| First Field Donation | 8.39 | 5.18 | 8.45 | 5.16 | $p=0.20$ |
| Individual contribution rate | 0.76 | 0.34 | 0.77 | 0.33 | $p=0.09$ |
| Average Field Donation | 8.66 | 4.15 | 8.84 | 4.05 | $p=0.09$ |
|  |  |  |  |  |  |
| No. of observations | 14,884 |  |  |  |  |

Note: The table presents summary statistics for people who never participated in an experiment and who participated in an experiment at least once. Panel A reports observable characteristics including the the person at registration, the number of semesters for which we observe donations, the individual's gende foreigner status, and the individual's field of study. Panel B summarizes our measures for prosocial beh "Contributed in first decision" is unity if the individual contributed to at least one of the two charitable fu his very first decision and zero otherwise. "First field donation" is the amount donated in the very first ded "Individual contribution rate" is the fraction of all possible decision in which the individual contributed least one of the two funds. "Average field donation" is the average amount that the individual donated in decisions.
${ }^{1} \chi^{2}$-tests for categorical variables and $t$-tests otherwise.
Table 2. Participating in Experiments Depending on Prosocial Behavior in the Field

| Dependent Variable | Participating at least once in a laboratory experiment |  |  |  |  |  | \# of Experiments participated in |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | (1) | (2) | (3) | (4) | (5) | (6) | (7) | (8) |
| Panel A: General Selection |  |  |  |  |  |  |  |  |
| First Field Donation | 0.001 | 0.002 | 0.002 |  |  |  | 0.002 |  |
|  | (0.003) | (0.003) | (0.003) |  |  |  | (0.013) |  |
|  | [0.000] | [0.000] | [0.000] |  |  |  |  |  |
| Avrg. Field Donation |  |  |  | 0.005 | 0.005 | 0.007* |  | 0.022 |
|  |  |  |  | (0.003) | (0.003) | (0.003) |  | (0.016) |
|  |  |  |  | [0.001] | [0.001] | [0.001] |  |  |
| Demographic controls | No | Yes | Yes | No | Yes | Yes | No | No |
| Field of study controls | No | No | Yes | No | No | Yes | No | No |
| Constant | -1.252** | -0.995** | -1.272** | -1.289** | -1.018** | -1.313** | -6.347** | -6.526** |
|  | (0.025) | (0.141) | (0.140) | (0.031) | (0.141) | (0.141) | (0.179) | (0.202) |
| No. of observations | 16,666 | 16,666 | 16,666 | 16,666 | 16,666 | 16,666 | 16,666 | 16,666 |
| Pseudo R squared | 0.00 | 0.023 | 0.050 | 0.000 | 0.024 | 0.050 | 0.000 | 0.000 |
| Mean of DV | 0.107 | 0.107 | 0.107 | 0.107 | 0.107 | 0.107 | 0.275 | 0.275 |
| Panel B: Subgroups |  |  |  |  |  |  |  |  |
|  | Female | Male | Econ | Non-Econ | Law | Arts |  |  |
| Avrg. Field Donation | 0.002 | 0.009 | 0.008 | 0.005 | 0.006 | 0.015** |  |  |
|  | (0.004) | (0.005) | (0.008) | (0.003) | (0.007) | (0.006) |  |  |
|  | [0.000] | [0.002] | [0.002] | [0.001] | [0.001] | [0.002] |  |  |
| Constant | -1.294** | -1.280** | -1.238** | $-1.301^{* *}$ | $-1.066^{* *}$ | -1.566** |  |  |
|  | (0.040) | (0.048) | (0.076) | (0.034) | (0.062) | (0.056) |  |  |
| No. of observations | 9,425 | 7,242 | 2,104 | 14,563 | 2,841 | 7,506 |  |  |
| Pseudo R squared | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 | 0.00 |  |  |
| Mean of DV | 0.101 | 0.115 | 0.120 | 0.105 | 0.155 | 0.077 |  |  |

[^11]Table 3. First Mover (FM) and Second Mover (SM) Behavior in Field Trust Game

| Dependent variable | FM Transfer |  | FM Belief |  | SM Back Transfer |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
|  | $(1)$ | $(2)$ | $(3)$ | $(4)$ | $(5)$ | $(6)$ |
|  | 0.299 | -1.486 | 0.821 | 0.588 | $-2.297^{* *}$ | -0.118 |
| Student | $(0.611)$ | $(0.797)$ | $(0.977)$ | $(1.467)$ | $(0.483)$ | $(0.904)$ |
|  |  |  | $1.502^{* *}$ | $1.445^{* *}$ | $1.597^{* *}$ | $1.623^{* *}$ |
| FM transfer |  |  | $(0.053)$ | $(0.062)$ | $(0.036)$ | $(0.039)$ |
|  |  |  | -0.019 | 0.026 | -0.056 | -0.062 |
| Student x FM transfer |  |  | $(0.108)$ | $(0.115)$ | $(0.067)$ | $(0.070)$ |
|  | No | Yes | No | Yes | No | Yes |
| Socio-demographic controls | $13.17^{* *}$ | $5.862^{*}$ | $-2.675^{* *}$ | -0.931 | $2.907^{* *}$ | -6.602 |
| Constant | $(0.287)$ | $(2.589)$ | $(0.452)$ | $(3.302)$ | $(0.285)$ | $(3.779)$ |
|  | 652 | 583 | 652 | 583 | 7,076 | 6,144 |
| No. of observations | 0.000 | 0.178 | 0.586 | 0.593 | 0.488 | 0.527 |
| R squared |  |  |  |  |  |  |

Note: The table investigates differences in first and second mover behavior in the trust experiment of study 2 . Columns (1) and (2) report OLS-estimations with average first mover transfers as the dependent variable (robust standard errors in parantheses). Columns (3) and (4) report OLS-estimations with average expected back transfers of first movers as dependent variable (robust standard errors in parantheses). Column (5) and (6) report OLS-estimations with second mover repayments as the dependent variable (robust standard errors clustered on individual in parantheses). As repayment decisions are elicited with the contingent response method, we have eleven observations per second mover (one for each possible first mover transfer). "Student" is an indicator variable which is one if the individual has been recruited from the student subject pool and zero otherwise. "FM transfer" is the first mover transfer. "Student x FM transfer" is the interaction effect of the two. Socio-economic controls include gender, age (and age squared), being an only child, being foreigner, being married, having obtained the general qualification for entrance to university or technical college, and dummies for political right- and left-wingers. Full estimation results can be found in Table A5 in the Web Appendix.
Level of significance: ** $\mathrm{p}<0.01,{ }^{*} 0.01<\mathrm{p}<0.05$


[^0]:    *We thank the editor, three anonymous referees, Simon Gächter, Lorenz Götte, Bruce Kogut, Rafael Lalive, and John List for helpful comments.
    ${ }^{\dagger}$ University of Bonn, Department of Economics, Adenauerallee 24-42, D-53113 Bonn; armin.falk@uni-bonn.de.
    ${ }^{\ddagger}$ Columbia University, Graduate School of Business, 710 Uris Hall, 3022 Broadway, New York, NY 10027; sm3087@columbia.edu.
    §University of Lausanne, Faculty of Business and Economics, Quartier UNIL-Dorigny, Internef 612, CH-1015 Lausanne; christian.zehnder@unil.ch.

[^1]:    ${ }^{1}$ Levitt and List (2007) and List (2009) focus on the latter possibility when they argue that behavior in the lab might not be a good indicator of behavior in the field.

[^2]:    ${ }^{2}$ As researchers we got access to the data through the university administration under the condition that we immediately anonymize the data after matching it with the experimental data base.

[^3]:    ${ }^{3}$ As robustness checks we also add estimations relying on a measure that counts how often individuals have contributed to at least one of the two funds. This Individual contribution rate correlates highly with the Average Field Donation (Spearman's Rho $=$ $0.92, p<0.001)$.

[^4]:    ${ }^{4}$ For results using contribution to at least one fund, see Web Appendix Table A2.

[^5]:    ${ }^{5}$ Cleave et al. (2010) use second mover back transfer in percent of the tripled first mover investment in a trust game as their measure of social preferences. On average second movers return about $25 \%$. They find that a one percentage point increase in the percentage returned decreases the participation rate by 0.09 percentage points. This is insignificant and the $95 \%$ confidence interval is $[-0.25,0.06]$ (in percentage points).
    ${ }^{6}$ We thank John List for pointing out this second margin of interest to us.

[^6]:    ${ }^{7}$ See Web Appendix Table A3 for the corresponding regressions with the number of experiments an individual participated in as dependent variable. Adding controls does not change the results.

[^7]:    ${ }^{8}$ First movers were also asked to indicate their expectation about the back transfer of their second mover given their own transfer decision.

[^8]:    ${ }^{9}$ All our results are robust if we use Tobit estimates to account for censoring.

[^9]:    ${ }^{10}$ Controls variables are gender, age (and age squared), being an only child, being foreigner, being married, having obtained the general qualification for entrance to university or technical college, and political opinions. Full estimation results can be found in Table A5 in the Web Appendix.

[^10]:    ${ }^{11}$ Table A5 in the Web Appendix reveals that being married and being a political leftwinger significantly increase second mover repayments.

[^11]:    Note: Columns (1)-(6) report Probit regressions where the dependent variable is 1 if the subject participated at least once in a laboratory experiment and 0 otherwise. Column (7)-(8) show Tobit regression where the dependent variable is the number of experiments participated in. Standard errors in parentheses. Marginal effects in brackets. Panel A shows regressions based on the whole population. Demographic controls are 'Age at registration', 'Number of semesters', dummies for 'Gender' and'Foreigner', and cohort dummies for the semester/year in which students registered. Field of study controls for the faculty of a person's major ( 'Computer Science', 'Economics \& Business', 'Theology', 'Law', 'Medicine', 'Veterinary medicine', 'Arts', and 'Natural Science'). The full regression results of the regressions are reproduced in Table A1 in the Web Appendix. Panel B shows coefficients of regressions for different subgroups. Level of significance: ${ }^{* *} p<0.01,{ }^{*} 0.01<p<0.05$.

