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Abstract

Group compensation and public announcement of performance are two common aspects of working with other people. We randomly assign these aspects to real-effort tasks. Following task completion and payment, subjects are given an unexpected opportunity to donate to a local charity. Group compensation and public announcement of performance have little effect on work performance but striking spillover effects on subsequent donations. Public announcement of performance doubles the amount donated to charity, and group compensation significantly increases the share donating. The results suggest that interpersonal interactions in the workplace environment may have important spillover effects on prosocial behavior outside of work.

JEL-Codes: D010, D640, A130.

Keywords: prosocial, spillover, charitable, group, experiment.

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1 Introduction

Working life can involve a variety of interpersonal interactions. Field evidence demonstrates multiple ways in which social interactions in the workplace affect worker productivity (Mas and Moretti, 2009; Bandiera et al., 2010; Chiaburu and Harrison, 2008). In this study we explore whether interactions during work have spillover effects on an individual’s behavior toward those outside of the workplace. We examine two common aspects of working with others: group compensation (i.e. interdependent payoffs, as occur with team projects, profit sharing, or bonuses) and public announcement of performance.

Long literatures in social psychology and experimental economics have advanced our understanding of the effects of groups. The “Social Identity Theory” of Tajfel (1970) and Tajfel et al. (1971) has endured over years of tests that have repeatedly shown that the mere perception of belonging to a group can alter behavior. Subjects can be categorized in groups described as “minimal” in that they are based on trivial similarities or even randomly assigned, yet this mere categorization has reliably induced in-group favoritism and out-group discrimination (Tajfel and Turner, 1986).

We hypothesize that working with others can affect prosociality toward those who are neither in the group nor even in the workplace. At least two theories would suggest such behavioral spillovers: social norms and group identity. First, prosociality may be motivated by adherence to social norms, and greater attention to others in a group could influence the salience of or beliefs about their behavior toward outsiders. In support of such norms, Bicchieri and Xiao (2009) found that manipulating expectations about labmates’ generosity influences subjects’ own generosity in the Dictator Game. Second, identifying with a group may provide psychological benefits (Akerlof and Kranton, 2000). Such benefits could affect attitudes and thereby have spillover effects. Indeed, Hargreaves Heap and Zizzo (2009) provided support for psychological benefits of artificial group while also indicating that these negatively affected trust toward labmates who were not in the group. Such spillovers need not be restricted to others in the laboratory, and broader spillovers should be expected if working with others affects brain chemistry. Social interactions can naturally enhance levels of chemicals such as oxytocin, which has been found to increase generosity toward other participants in the laboratory and toward real charitable organizations (Zak et al., 2007; Barraza and Zak, 2009; Barraza et al., 2011; Van IJzendoorn et al., 2011).

This study employed a laboratory experiment designed to test for prosociality spillovers of working with others. All subjects performed the same set of real-effort tasks for compensation. Randomization followed a two-by-two design that allowed us to estimate isolated effects of two common

aspects of working with others as well as any potential interaction between these aspects. First, some subjects were paid a piece rate based on their own performance (“Individual”), while others were randomly assigned to groups and paid based on the group’s performance (“Group”). Second, some subjects’ performance was kept as private knowledge (“Private”), while others were informed that at the end of the tasks they would be asked to announce their performance to everyone in the laboratory (“Public”). The two-by-two design examined the “Group” and “Public” treatments together and in isolation, reflecting the imperfect correlation across workplaces between these aspects of working with others. Regardless of treatment group, subjects were told they would be paid for correct answers provided in each of four tasks. Subjects were not informed before the tasks that they would later have an opportunity to donate from their earnings. Upon completion of the tasks, subjects were shown a screen stating that they had completed all four tasks, then shown the amount they had earned, then given an opportunity to donate a portion of their earnings to a local charity, then asked to complete a survey, and then paid. Reflecting real-life working environments, our experimental design is tailored to explore the spillover effects on pro-social behaviour allowing for interactions of both aspects of working with others because these are not perfectly correlated across workplaces.

We found that both treatments significantly affected donations despite having no significant effects on performance or earnings. Specifically, each treatment increased donations, with the result obtaining statistical significance on the share donating for the “Group” treatment, on the amount donated for the “Public” treatment, and in Tobit regressions for both treatments. The pattern of results is consistent across levels of individuals’ earnings, group earnings, and subject scores on the Cognitive-Reflective Test of Frederick (2005). We posit that these effects are likely driven by the social treatments’ manipulation of the extent to which subjects think about other individuals besides themselves. Consistent with this other-regarding mechanism, we find that subjects who report knowing other subjects in the laboratory show significantly stronger donation responses to the social treatments despite being no more generous than other subjects in the baseline “Individual, Private” treatment.

Our study adds to existing literature looking at the role of behavioral spillovers (see, e.g., Dolan and Galizzi, 2015; McCannon and Rodriguez, 2016). Bruhin et al. (2016) find evidence of *interpersonal* spillovers in blood donation, whereas we examine an *intrapersonal* spillover from the earnings environment to the donation decision. Pan and Houser (2013) found that subjects assigned to groups that were given a more cooperative work task were subsequently more generous toward out-group members in trust games, a result similar to the spillovers we find. The direction of causal-

ity of our spillover is opposite that found by Anik et al. (2013), who provide evidence that prosocial bonuses in the form of donations to charity lead to happier and more satisfied employees at an Australian bank. Our experiment is more similar in design to those of Erkal et al. (2011) in that subjects earn money based on performance and can then give some away, particularly their variant in which subjects are not told about donation opportunities until after they earn. Erkal et al. (2011) induce sizable earnings differences between each group member and study gifts to members of the group, whereas we pay all subjects a piece rate and study donations to a real charity.¹

Our findings offer a novel consideration for interpreting studies of audience effects on prosociality. These studies have shown that identification of subjects and their decisions through announcements or the presence of an audience may have a positive impact on charitable donations (Ariely et al., 2009), establishing a 50-50 sharing norm (Andreoni and Bernheim, 2009) and increasing contributions to a public good (Andreoni and Petrie, 2004; Rege and Telle, 2004; Soetevent, 2005; Alpizar et al., 2008; Denant-Boemont et al., 2011; Samek and Sheremeta, 2014). Whereas these studies expose the prosocial behavior to an audience, our study exposes work performance to an audience. The fact that we still find significant spillover effects on private prosocial behavior isolates direct effects of exposure to an audience that may be operating in addition to the social image concerns that are generally thought to mediate audience effects on prosociality.

The effects of the aspects of group work that we study may be viewed as an additional example of studies that have found effects of priming with social factors and emotional state on prosociality. Inducing anger among subjects lowers contributions in a public goods game (Drouvelis and Grosskopf, 2016), while priming subjects on cooperation increases contributions (Drouvelis et al., 2015). Priming subjects with religious words has increased donations in the Dictator Game (Shariff and Noranzayan, 2007). Priming subjects with a picture of eyes has been found in the laboratory to increase donations in the Dictator Game (Haley and Fessler, 2005) and in the field to increase payments for purchases (Bateson et al., 2006), decrease littering (Ernest-Jones et al., 2011), and increase recycling (Ekström, 2012). Allowing subjects to identify with their experimental partners by viewing them (Bohnet and Frey, 1999), seeing photographs of them (Scharlemann et al., 2001), or communicating with them (Bochet et al., 2006) has been shown to generate positive effects on prosocial behavior. Putting givers “in the receiver’s shoes” by having subjects make decisions for both roles before randomly assigning them to either the giver or receiver role had a significant positive effect on the amount given (Andreoni and Rao, 2011). More generally, this study contributes to

¹Our paper takes up the challenge in theirs that “Future work can consider whether similar results hold in other environments, such as those where earnings are determined using a piece-rate scheme and where subjects give to an actual charity instead of each other.”

a growing body of experimental literature exploring the behavioral determinants of charitable donations. Many of these studies have examined the importance of economic mechanisms such as using matching to alter the price of giving (Karlan and List, 2007) and providing gifts for potential donors (Falk, 2007; Garbarino et al., 2013). Increasingly, however, this research highlights the importance of social interactions in determining charitable giving.

The paper proceeds as follows. Section 2 describes the design of the experiment. Sections 3 and 4 present the results for earnings performance and for donations, respectively. We discuss the findings in Section 5 and conclude in Section 6.

2 Experiment Design

To address our research questions, we employed a 2x2 factorial (between-subjects) design analyzing the effects of two different incentive schemes (individual piece-rate vs. revenue-sharing forms of compensation) and information structures (public announcement vs. no public announcement). We measure productivity by observing subjects' performance in four different timed tasks. Similar to Wozniak et al. (2014), our design consists of two types of tasks: math and word tasks. Having each subject complete both types of tasks allows all subjects to respond to treatment even if, say, certain cultural groups or genders have difficulty with one type of task (Niederle and Vesterlund, 2010). In addition, each of these two tasks consists of an easy and hard version. Subjects were given a 5-minute time limit for each task and were shown the timer.

For both the math and the language task, items were presented to subjects on a computer screen. Subjects would type in an answer and click the "Submit" button. After each submission, a new item was immediately shown. For the math task, subjects were asked to multiply two numbers. The easy (hard) version of this task involved the multiplication of two one-digit (two-digit) numbers. For the word task, each subject had to arrange pairs of letters to form a word. The easy (hard) version of this task involved re-arranging two (four) pairs of letters. Subjects were told that they have to use all pairs of letters to form the correct word and can re-arrange the order of the pairs but not the order of the letter within each pair. For both the math and the word task, two sheets of scratch paper and a pen were provided, but no other form of assistance was available.

Each subject performed all four tasks. At the beginning of each task, instructions were provided on screen and read aloud. Subjects within a session performed tasks of equivalent difficulty at the same time, and we alternated the order of the tasks across sessions.² Prior to any of the four tasks,

²Specifically, we ordered the more familiar math task first but varied the order in which the easy or the hard version

subjects were provided with written and read instructions that described the general form of the treatment. The general instructions varied according to the two dimensions of treatment: whether subjects were compensated in groups and whether performance was announced.

We study two incentive schemes: a piece-rate and a revenue-sharing compensation scheme. In treatments with the piece-rate compensation (“Individual” treatments), subjects were paid according to their own performance. This means that the payoff a subject receives is equal to the piece rate multiplied by the number of correct responses of the subject for that particular task. The piece-rate incentives differ depending on whether subjects perform the easy or the hard version of a task. In both the math and the word tasks, a subject received £0.03 for each correct answer in the easy task and £0.21 for each correct answer in the hard task. As the hard version of a task requires subjects to put in more effort to provide a correct answer relative to the corresponding easy version, we chose a higher compensation rate for the former compared to the latter version of a given task.

In treatments with the revenue-sharing form of compensation (“Group” treatments), subjects were randomly assigned to 3-person groups and were paid according to the group’s performance. We maintained the same piece rate of £0.03 (£0.21) for each easy (hard) answered correctly, but subjects were paid the piece rate multiplied by the average number of correct responses in the group for that particular task. In other words, the payments for a group’s total productivity were split evenly among all 3 members of a group. Subjects were not told which subjects were in their group but were given the number of their group as a label.

The second dimension of treatment varied the information structure. In half of the sessions (“Private” treatments), subjects learned about only their own performance. In the other half of the sessions (the “Public Announcement” or “Public” treatments), subjects were informed that at the end of final task they would be asked to stand up and tell the others in the room how many correct answers they had provided in each of the four tasks. The “public” treatment was therefore similar to the group announcement and audience treatments in past work, though the announcement related to performance and not to the donations outcome of primary interest.³ Importantly, subjects received feedback about performance only after completion of the final task.⁴

After completion of the final task, subjects were informed of their earnings from the experiment. They were then shown a screen asking whether they would like to donate part of their earnings to

of each task was experienced first. Subjects therefore performed the tasks in one of two orders: [easy math, hard math, easy word, hard word] or [hard math, easy math, hard word, easy word]. We find no effect of the ordering on earnings or donations.

³This and all other treatments received ethics approval from the Institutional Review Board. Subjects were allowed to withdraw from the experiment at any point.

⁴At this time, subjects in the “Group” treatments also received feedback about the average number of correct responses in their own group.

a well-known local charity.⁵ Subjects could obtain information about the charity by clicking on its onscreen logo. An important feature of our design is that subjects were not informed about the option to donate to the charitable cause until after they had completed all tasks and, in relevant treatments, learned about the performance of the other subjects in their group or their session. This aspect of our design enables us to analyze potential spillover effects on charitable giving which may come from the form of compensation scheme, information sharing, or the interaction of both. After subjects’ donation decision had been made, we also collected data on their demographic characteristics and administered the Cognitive Reflection Test (CRT) of Frederick (2005). The CRT is a short list of questions, three of which have an answer that is seemingly intuitive but incorrect, and correct answers to these questions provide a measure of one’s proclivity for reflection that is correlated with cognitive outcomes.

We conducted 16 sessions: two sessions for each of the two sequences and each of the four treatment combinations. A total of 282 subjects participated in the experiment. A breakdown of the number of participants per sequence and treatment is given in Table 1.

Table 1: Overview of experimental treatments

	Private	Public
Individual	Sequence 1: 34 subjects Sequence 2: 36 subjects	Sequence 1: 36 subjects Sequence 2: 35 subjects
Group	Sequence 1: 36 subjects Sequence 2: 36 subjects	Sequence 1: 33 subjects Sequence 2: 36 subjects

All the subjects were recruited at the University of Birmingham, using the ORSEE software (Greiner, 2015) and were randomly selected from a large database of volunteers who have previously signed up to participate in economics experiments. The vast majority of participants were undergraduate students from various academic fields. Subjects were allocated to treatments at random. The experiment was conducted in the Birmingham Experimental Economics Laboratory (BEEL) and all treatments were computerized and programmed with the Multistage software from Caltech. The full set of instructions used in the experiment is provided in Appendix A. At the end of a session, subjects were paid in private according to their total earnings from all relevant tasks. Average earnings per treatment (including a show-up fee of £2.50) were as follows: £15.11 for the “Individual, Private”, £16.11 for the “Group, Private”, £16.17 for the “Individual, Public”, and £15.34 for the “Group, Public.”⁶ Sessions lasted, on average, 65 minutes. Appendix Table B.1

⁵The charity is called Acorns Children’s Hospice of Birmingham, which provides specialist care for babies, children and young people who are life-limited or life-threatened. More information about the charity can be found here: <https://www.acorns.org.uk/>.

⁶At the time of the experiment £1 was equivalent to US\$1.52.

shows that treatment groups are balanced in terms of pre-determined subject characteristics captured by survey. Appendix Table B.2 shows a descriptive analysis of these pre-determined subject characteristics; making a donation is positively correlated with age, being married, and having a non-UK nationality.

3 Analysis of Performance

We first examine whether the treatments affected subjects’ earnings performance. Subjects were paid for correct answers to each of four tasks. We provide summary statistics showing that subjects were able to perform all four tasks and that there was variation in the number of each task completed. We then focus on earnings from a subject’s own performance in these tasks as a summary measure of performance in the experiment.

Table 2 presents subjects’ average performance as measured by the number of correct responses provided for each of the tasks and treatments. As anticipated, we observe that subjects were more productive in the easy tasks than in the difficult ones. In all four treatments, the best performance is recorded in the easy math task and the worst performance is observed in the hard math task. The coefficient of variation is greater than 0.15 in all cases and nearly 0.7 in some, indicating considerable variation in performance within all tasks. Appendix Table B.3 displays regressions testing for significant differences in the number of responses attempted, left blank, and answered correctly. Estimated effects of the “Group” and “Public” treatments on each of these three outcomes are generally positive and not statistically significant.

Table 2: Average Correct Responses by Task and Treatment

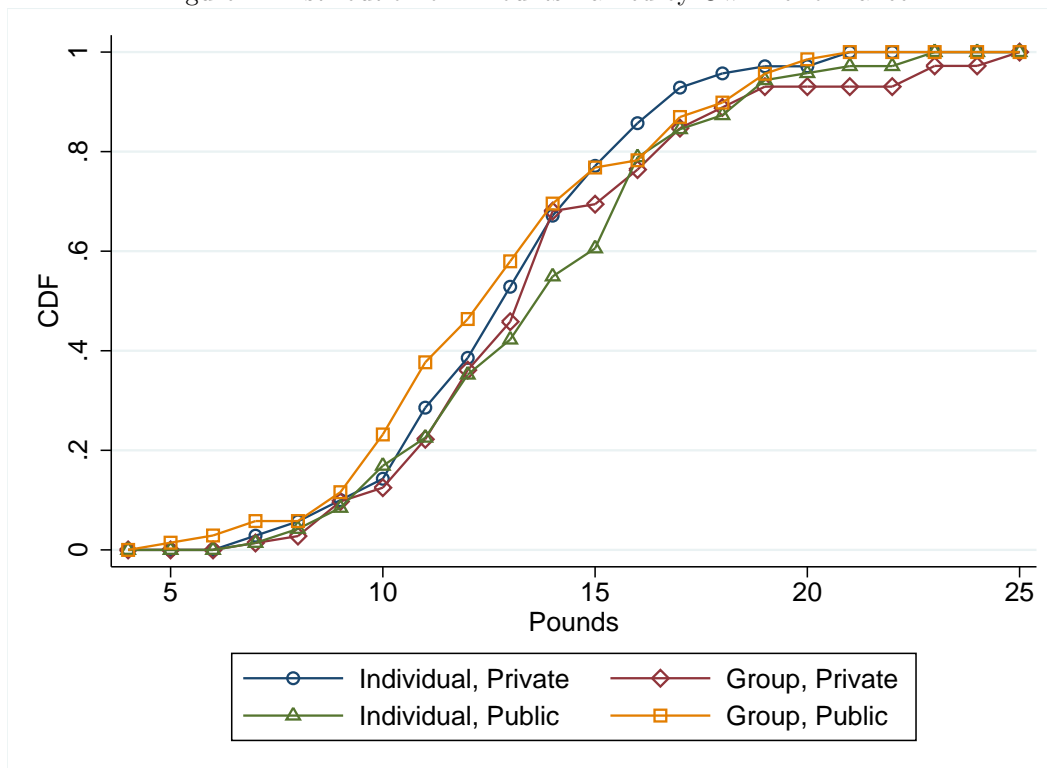
	Private				Public			
	Math		Word		Math		Word	
	Easy	Hard	Easy	Hard	Easy	Hard	Easy	Hard
Individual	150.24 (40.14)	8.64 (6.00)	99.09 (18.69)	16.89 (7.37)	154.89 (44.22)	9.24 (5.64)	104.04 (16.17)	18.86 (8.34)
Group	150.92 (45.33)	8.65 (5.47)	103.97 (17.07)	19.75 (9.18)	144.78 (48.81)	7.22 (4.89)	98.48 (19.00)	18.07 (7.48)

Numbers in the cells indicate average performance (number of correct answers) in each task. Numbers in parentheses indicate standard deviations.

We use earnings from a subject’s own correct answers as a summary measure of performance. Earnings from own performance, or own earnings, is defined as the total earnings generated by the subject, whether or not these were split with a group, so that that the measure is consistent across

the “Individual” and “Group” treatments.⁷ Figure 1 displays the cumulative distribution function (CDF) of this measure of earnings from one’s own performance. Subjects in the “Group, Public” treatment were slightly more likely than others to earn amounts close to £10, while those in the “Individual, Private” treatment were slightly less likely to earn amounts greater than about £17, but overall the distributions appear similar.

Figure 1: Distribution of Amounts Earned by Own Performance



Notes: N=282. Kruskal-Wallis equality of populations rank test p-value=0.261.

To test for differences in performance across treatments we estimate OLS regressions of the form

$$Y_i = \beta_0 + \beta_1 \cdot 1[\text{Group, Private}_i] + \beta_2 \cdot 1[\text{Individual, Public}_i] + \beta_3 \cdot 1[\text{Group, Public}_i] + \epsilon_i \quad (1)$$

In equation 1 the independent variables are indicators for each of the treatments other than the baseline “Individual, Private” treatment. Thus, β_0 gives the average value of the outcome in the baseline treatment, and the other coefficients each show the difference between this and the average in one of the other treatment arms. For most regressions we report heteroskedasticity-robust standard

⁷Recall that the piece rates paid to the group in “Group” treatments were the same as the rates paid to the subject in “Individual” treatments.

errors, which are equivalent to standard errors clustered by subject because each outcome is only measured once per subject. When we explore the effect of group earnings we cluster by group to account for the correlation in earnings among group members.

We report the estimates from our own-earnings regressions in Table 3. The dependent variable is earnings from one’s own performance, as previously described and shown in Figure 1. The main results appear in model (1), in which we include the entire experimental sample. The first two coefficients indicate that both the “Group” and “Public” treatments in isolation increase earnings by about £0.80 over the mean of £12.84 in the control group, but neither effect is statistically significant. The point estimate for the effect of combining the treatments is even smaller.

Table 3: Effects on Own Earnings, in Aggregate and by Cognitive Level

	All Subjects	CRT=0	CRT=1	CRT=2	CRT=3
	(1)	(2)	(3)	(4)	(5)
Group, Private	0.770 (0.573)	-0.032 (1.077)	3.147*** (1.041)	-0.700 (0.891)	-1.266 (1.471)
Individual, Public	0.828 (0.538)	-0.151 (0.809)	2.118** (1.034)	-0.124 (1.000)	0.703 (1.209)
Group, Public	-0.232 (0.553)	-1.002 (0.876)	0.966 (1.089)	-3.180*** (0.726)	1.777 (1.108)
Constant	12.841*** (0.349)	12.194*** (0.589)	11.340*** (0.556)	14.850*** (0.561)	14.445*** (0.780)
N	282	93	75	66	48
Adj. R-Squared	0.007	-0.015	0.077	0.128	0.036

Notes: Heteroskedasticity-robust standard errors are in parentheses. In all models the dependent variable is an individual’s own earnings in a given task.

To explore the performance findings further we also report the estimates when the sample is restricted to individuals receiving different CRT scores. Recall that we assessed one type of cognitive ability using subjects’ responses to the Cognitive Reflection Test introduced by Frederick (2005). The research suggests that subjects with a higher number of correct responses are more likely to reflect on their answers and less likely to choose intuitive responses. In Models (2)-(5) of Table 3 we report results from four separate regressions corresponding to each possible number of correct responses to the CRT task. When we look separately at each sub-sample of subjects, we find that the pattern of the main result in Model (1) is driven almost entirely by subjects who provided only one correct answer to the three CRT questions (Model (3)). The small value of the constant in this column shows that these subjects performed worse than all others in the control group, including those with CRT equal to zero (though the difference is not statistically significant). These individuals performed

significantly better when either compensation depended on group performance or performance was announced publicly, but not when both of these held. By contrast, we find nearly opposite patterns when we look at subjects with higher cognitive abilities. Individuals with a CRT score of either 2 or 3 perform significantly better than others in the baseline treatment, but the insignificant point estimates for the “Group” treatment show that they earned roughly £1 less when sharing these earnings with a group. We interpret these results as suggesting that while high-ability subjects may free-ride, low-ability subjects may exert more effort when they know their performance will be shared with others, perhaps to save face or avoid letting the group down.

Table 3 showed that average effects on earnings were not significant. Appendix Table B.4 shows that effects on the distribution of income were also limited. The “Public” treatment only increased the earnings of relatively high performers, and “Group” treatment effects were all statistically insignificant.

While our main focus is donations and not performance, for context we offer some potential explanations for the lack of significant effects on performance. First, subjects’ effort in the tasks may have been determined by intrinsic motivation, such as a desire to see how many correct responses they could provide. Such intrinsic motivation could weaken the effects of extrinsic motives (see Gneezy et al. (2011) for a review on motivation crowding-out effects). Second, Corgnet et al. (2015) document free-riding by Internet browsing, but subjects in this experiment had no access to the Internet or other alternative use for their time and hence could not fully benefit from free-riding. Third, the treatments may have had offsetting effects on motivation, such as if the group treatment reduced financial incentives but induced altruism or competitiveness toward group members. Assessing the importance of such factors may be useful for research focused on work incentives and performance measurement.

4 Results for Charitable Donations

In this section we examine our primary outcomes: the spillover effects from group compensation and announcement of performance onto subsequent prosocial behavior. We reiterate that these are, in fact, spillover effects, because donations were made privately, after performance and compensation, in a manner that was not directly affected by the treatments. We first report summary statistics on donations in each treatment and show the CDF for each. We then estimate the effects of treatment on the extensive and intensive margins of donations, and we perform heterogeneity analysis to describe the most likely mechanism for these effects.

Average donations and standard deviations are reported in Table 4. In all four treatments, the majority of subjects are giving a positive amount to the charity. In particular, 51% of the subjects do so in the “Individual, Private”, 62% of the subjects in the “Individual, Public”, 72% of the subjects in the “Group, Private” and 71% of the subjects in the “Group, Public”. This provides evidence that, on average, subjects are willing to sacrifice part of their earned income from the experiment. Across all subjects, average donations are highest in the “Individual, Public” treatment (£1.42) and lowest in the “Individual, Private” treatment (£0.67).

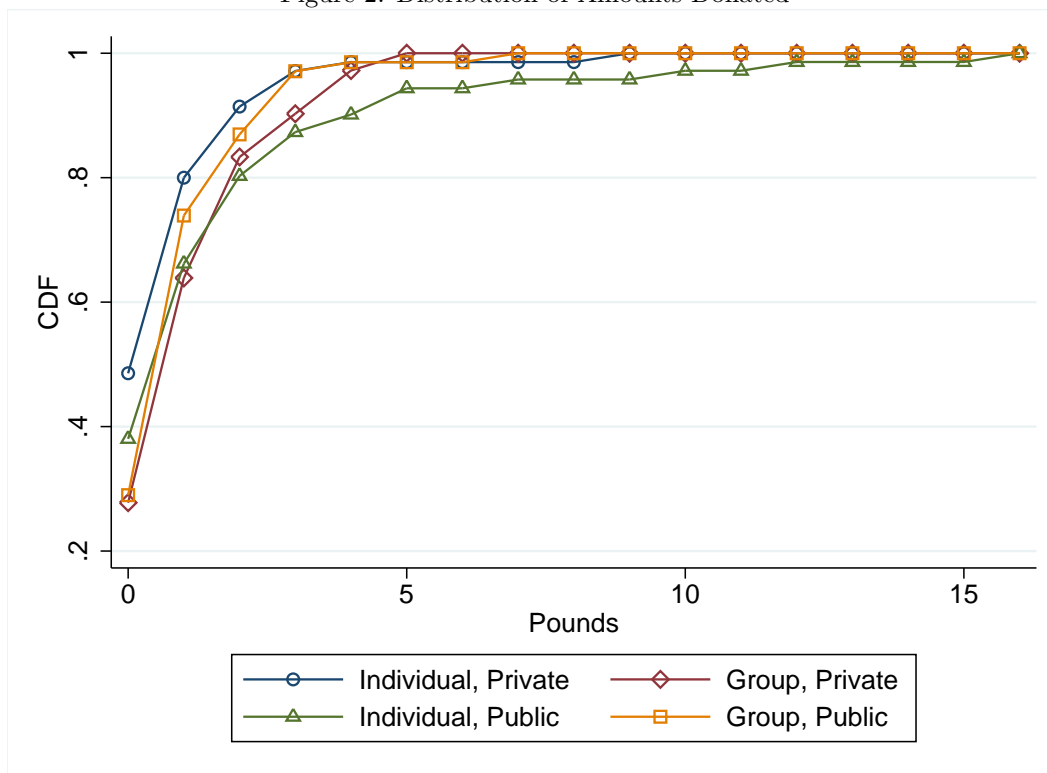
Table 4: Average Donations

	Private	Public	Total
Individual	0.665 (1.316)	1.423 (2.724)	1.047 (2.170)
Group	0.983 (1.201)	0.859 (1.092)	0.922 (1.146)
Total	0.827 (1.265)	1.145 (2.097)	0.985 (1.733)

Notes: Numbers in cells correspond to average donations (in pounds). Numbers in parentheses correspond to standard deviations.

Figure 2 displays the CDF for each treatment. Mostly clearly we see the differences on the extensive margin of choosing to make a nonzero donation, with subjects that were compensated in groups showing a greater likelihood of doing so. While it is the group treatments that appear to have the largest effect on the extensive margin of giving or not, the “Individual, Public” treatment appears to have had the greatest effect on the intensive margin of amount donated. We can also see that the distribution of gifts in the baseline “Individual, Private” treatment is nearly dominated by the distributions of the other three treatments, and that “Group, Public” is nearly dominated by the other two. Hence, as with performance, the “Group” and “Public” treatments appear to have the greatest effect when used in isolation.

Figure 2: Distribution of Amounts Donated



Notes: N=282. Kruskal-Wallis equality of populations rank test p-value=0.0676.

The main results of the paper are displayed in Table 5. We formally test for treatment effects on donations by again estimating equation 1. Here we examine as outcomes a dummy variable indicating that an individual gives to the charity (in Model (1)), the unconditional amount donated (in Model (2)), and the amount donated conditional on making a nonzero donation (in Model (3)). We also estimate a Tobit model to simultaneously account for responses on both the extensive and intensive margins (in Model (4)). Focusing first on the more comprehensive Tobit model, we find that both the “Group” and “Public” treatments significantly increase prosociality when used in isolation, and that the effect of “Group, Public” is also positive and marginally significant.⁸ Turning to the linear models, we see the same pattern of coefficients. Specifically, Model (1) indicates that subjects are significantly more likely to donate if they were compensated in groups rather than individually. The point estimate of the effect of public announcement alone was positive but half as large and not statistically significant. Model (2) indicates that announcing performance in public did significantly increase the amount donated under individual compensation, by approximately £0.76,

⁸We can reject (with p-value 0.0314) the hypothesis that the “Group” and “Public” treatment dimensions have additive effects. We surmise that payoff commonality and public announcement of performance do not reinforce each others’ effects because the combination invites comparison of oneself with others in the group. In evidence of the negative feelings generated by such comparisons, Card et al. (2012) find that information about peers’ salaries causes those receiving below-median pay to report lower satisfaction and a higher prevalence of looking for a new job.

on average, more than doubling the baseline mean contribution of £0.67. This treatment also shows a positive and marginally significant effect on the amount conditional on a nonzero donation (Model 3).⁹ Overall, these results provide evidence that all treatments had significant positive spillover effects on prosocial behavior.

Table 5: Spillover effects on donations – Regression results

	(1)	(2)	(3)	(4)
	Any Donation	Amount	Cond. Amount	Tobit
Group, Private	0.208** (0.080)	0.318 (0.212)	0.068 (0.316)	0.810** (0.369)
Individual, Public	0.105 (0.084)	0.758** (0.359)	1.003* (0.547)	1.110** (0.535)
Group, Public	0.196** (0.082)	0.194 (0.205)	-0.084 (0.311)	0.649* (0.367)
Constant	0.514*** (0.060)	0.665*** (0.157)	1.293*** (0.267)	-0.296 (0.321)
N	282	282	181	282
Adj. R-Squared	0.020	0.015	0.034	

Heteroskedasticity-robust standard errors are in parentheses. Columns (1)-(3) report results of OLS regressions. Dependent variables, by model, are an indicator for a nonzero donation (1), the amount donated (2), and the amount donated conditional on a nonzero donation (3). Model (4) is a Tobit model for donations, and reported coefficients are marginal effects on observed donations.

We provide several additional tests to explore two potential mechanisms for prosociality spillovers: income effects and awareness of others. For one, to the extent that the treatments increased earnings, earnings could have increased charitable donations. Though the earnings effects reported in Table 2 were not statistically significant, the point estimates for the two main treatments were positively signed. We offer six pieces of evidence that the prosociality spillovers were not driven by income effects. First, the point estimate for the effect of the “Group, Public” treatment on earnings was negative, yet this treatment significantly increased the probability of a donation. Second, the point estimates for the “Individual, Public” treatment indicate that it raised earnings by £0.828 (6.44 percent) and donations by £0.758 (113.99 percent), implying an income elasticity of 17.7 (=113.99/6.44), which is an order of magnitude larger than income elasticities estimated in the literature.¹⁰ Third, as we show in Appendix Table B.5, the pattern of spillover effects is the same across all cognitive levels. When we drop subjects with only one correct answer to the CRT

⁹Difference in conditional mean suggest responsiveness on the intensive margin but should be interpreted with caution given responses on the extensive margin, which may involve new gifts of positive-but-small amounts that would lower the conditional mean.

¹⁰Observational studies have examined income effects on donations using individual income tax data. Auten et al. (2002) obtain estimates ranging from 0.29 to 0.45 for the elasticity of charitable donations with respect to transitory annual income. Similarly, the corresponding estimates of Bakija and Heim (2011) are close to 0.33. In a laboratory study, Erkal et al. (2011) find an earnings elasticity of giving to lab-mates of 1.2 when payoffs are randomly assigned.

questions, those whose earnings increased most under the treatments, we obtain noisier but larger estimated effects in the Tobit model estimated in Table 5. Fourth, we use random variation in earnings to show that the causal effect of income on donations was small. Specifically, we use the “Group, Private” and “Group, Public” treatment samples and regress donations on the amount of earnings that were generated by group members. Since groups were randomly assigned this provides random variation in income, and the regression results in Table 6 show that additional income had small and statistically insignificant effects on donations. Fifth, Appendix Table B.6 shows that donations are also not correlated with earnings due to the individual’s own performance or the interaction of these with earnings generated by the group.¹¹ Sixth, Appendix Table B.7 repeats the regressions above but include a control for the subjects’ total earnings (not included in our preferred specifications in Table 5 because income is endogenous), and results are unchanged. These results are consistent with the findings of both Drouvelis and Marx (mimeo) and Tonin and Vlassopoulos (2017) that income in experiments did not affect donations.

Table 6: Irrelevance of Earnings to Donation Decision

	All Group Sessions		Group, Private Sessions		Group, Public Sessions	
	(1) Donation	(2) Any Donation	(3) Donation	(4) Any Donation	(5) Donation	(6) Any Donation
Groupmembers’ Earnings	0.035 (0.045)	0.006 (0.021)	-0.044 (0.057)	0.023 (0.030)	0.125 (0.074)	-0.016 (0.030)
Constant	0.614 (0.387)	0.660*** (0.191)	1.384** (0.542)	0.512* (0.282)	-0.193 (0.589)	0.849*** (0.265)
N	141	141	72	72	69	69
Adj. R-Squared	-0.004	-0.007	-0.010	-0.006	0.020	-0.012

Heteroskedasticity-robust standard errors are in parentheses. In odd-numbered models the dependent variable is the amount donated to the charity. In even-numbered models the dependent variable is an indicator for a nonzero donation.

Finally, we explore heterogeneity along a dimension related to our hypothesis that the treatments increase thoughts about others. One of the questions subjects were asked in the survey was “How many participants in the experiment do you know by name?” We used this question to split the sample into subjects who knew at least one other person in the laboratory and subjects who knew no other participants. Subjects who know other subjects by name are possibly most representative of members of real work groups and are probably most likely to think about others when treated with a connection to others in the laboratory, especially when the connection is a public announcement to

¹¹Individuals’ own earnings are potentially endogenous to the donation decision, but their interaction with exogenous group earnings would capture effects of *relative* performance on donations.

everyone in the laboratory. We ran the same set of donation regressions on each of the subsamples, and the results appear in Table 7. The values of the constants in these regressions show that individuals who know others in the laboratory are no more generous than their peers in the baseline “Individual, Private” treatment. However, the estimated effects of each of the “Group” and “Public” treatments are greater among the sample of subjects who know others. In this subsample the effects are statistically significant, even at the 0.01 level for the “Individual, Public” treatment, despite the fact that it makes up only about one quarter of the full sample. Again, these results are not driven by income effects, as we show in Appendix Table B.8 that earnings in each treatment are quite similar between subjects who know others and subjects who do not. We take these results as evidence that thinking about others is an important mechanism behind the estimated prosociality spillovers of working with others.

Table 7: Importance of Social Connections to Donation Decision

	(1)	(2)	(3)	(4)	(5)	(6)
	Any Donation	Amount	Tobit	Any Donation	Amount	Tobit
Group, Private	0.171* (0.095)	0.113 (0.259)	0.526 (0.452)	0.306** (0.152)	0.827** (0.338)	1.394** (0.556)
Individual, Public	-0.009 (0.100)	0.597 (0.484)	0.690 (0.697)	0.388*** (0.146)	1.169*** (0.376)	1.836*** (0.565)
Group, Public	0.206** (0.092)	0.194 (0.262)	0.686 (0.447)	0.141 (0.172)	0.131 (0.238)	0.405 (0.533)
Constant	0.549*** (0.070)	0.764*** (0.205)	-0.186 (0.394)	0.421*** (0.116)	0.399** (0.171)	-0.434 (0.413)
N	204	204	204	78	78	78
Adj. R-Squared	0.027	-0.001		0.063	0.117	

Heteroskedasticity-robust standard errors are in parentheses. The sample is split into subjects who report knowing zero other subjects in the lab (models 1-3) and subjects who know at least one other subject in the lab (models 4-6). Dependent variables are either an indicator for a nonzero donation (models 1 and 4), or the amount donated. Models 3 and 6 are Tobit models for donations, and reported coefficients are marginal effects on observed donations.

5 Discussion

In this section we address three goals. The first is to consider the evidence for which mechanism drives the spillover effects we find. The second is to place these findings in the larger literatures on charitable giving and social capital. The third is to note limitations of the study.

We conducted this research because a variety of noted studies implied that working with others might have spillover effects on prosociality in other domains. Existing work suggests a variety of mechanisms by which such spillovers could occur. One possibility is the idea of “conscience

accounting,” as proposed by Gneezy et al. (2014), who found that behaving immorally increased subjects’ guilt and subsequent donations. In our setting, subjects in the social treatments might feel they could have done more in terms of performance, which might lead them to donate more. However, this explanation does not appear consistent with the full set of results. For one, the social treatments do not have a significant average effect on performance, and hence most subjects were likely working as hard as they would have in the control group.¹² For another, high-CRT subjects do appear to free-ride in terms of performance when in groups, yet the effect on their donations is similar to the effect for the average subject. It also seems unlikely that the primary mechanism is the psychological benefits of group membership proposed by Akerlof and Kranton (2000). While such effects may be relevant for the “Group” treatments, the “Public” treatments did not place subjects in groups, and these treatments had the greatest effects on the amount donated.

In our view, our work treatments most likely generate prosociality spillovers because they encourage subjects to think about individuals besides themselves. Past research has found that individuals become more generous when they are asked to put themselves in others shoes (Andreoni and Rao, 2011) or when they learn that others have been generous (Bicchieri and Xiao, 2009). Greater awareness of or attention to other individuals could therefore increase donations by making subjects think more about either the recipients of their gifts or about the likely donations of the peers whose presence has been made more salient. The fact that our effects are strongest among subjects who know other subjects in the laboratory suggests the latter channel, i.e. that subjects consider the likelihood of generosity by others and then attempt to adhere to this norm. This channel is supported by a finding of Drouvelis and Marx (mimeo) that when a subject knows that labmates have given an additional £1 per person she will herself give an additional £0.53 on average.

Our finding of prosociality spillovers fits into a growing body of economic research on social motivations for giving. For example, Meer (2011) found that college graduates were more generous toward their alma mater when socially connected to their solicitor. One result that has proven robust to numerous settings is that informing individuals about the donations of others tend to increase these individuals’ donations (List and Lucking-Reiley, 2002; Shang and Croson (2009); Huck and Rasul, 2011; Huck et al., 2015). Donations have also been shown to respond to expressions of support for the charity (Kessler, forthcoming), solicitor attractiveness (Landry et al., 2006; Landry et al., 2010), repetition of requests (Meer and Rosen, 2011), and social pressure (DellaVigna et al., 2012). The results of our experiment suggest that merely interacting with a random group of people can

¹²Appendix Table B.3 shows that for individual tasks there were some significant differences in the number of items left blank, but the sign of these differences is not consistent across Group treatments, and there are almost no significant differences in the number of correct responses.

affect the degree to which we look outwards and donate.

Our results also provide evidence that may also be relevant for the literature on “social capital,” the idea of the social structure as a resource (Coleman, 1988; Borjas, 1992). Putnam (2000) makes the case that social interaction has decreased over time and that this decrease may have negative sociological consequences. Consistent with the possibility of such consequences, trust and civic norms have been found to correlate with national economic performance (Knack and Keefer, 1997), and multiple measures of social capital are highly correlated with income mobility across U.S. counties (Chetty et al., 2014). Most arguments for economic effects of social capital emphasize reductions in transactions costs, but our results suggest that social interactions could also increase efficiency by inducing spillover effects on externalities outside of any particular network.

Finally, we note some important limitations of the study. The spillover effects we estimate in the controlled environment of the laboratory may differ from those occurring in real-world workplaces. In particular, we estimate the effects of two particular aspects common to working with others, but we have abstracted from other important aspects, such as social interplay, strategy, and hierarchy. These aspects of groups may have either positive or negative spillovers, and they may overwhelm the effects of the aspects we examine. We hope that our findings of prosociality spillovers from some aspects of working with others will motivate research into other aspects and working environments.

6 Conclusion

We have examined the impact of two aspects of group work on individuals’ work performance and prosocial behavior: compensation based on individual or group performance and performance that was either private or public information. We find that the studied aspects of working with others have little effect on performance but significant spillover effects. Specifically, the public announcement of performance more than doubles the amount donated to the charity. Group-based compensation significantly increases the probability of subjects’ donating to the charity. Each has significant effects in a Tobit regression. Our treatment effects follow similar patterns across subjects with different cognitive abilities and incomes but are strongest among subjects who know other subjects in the laboratory.

The spillover effects we estimate have relevance for manager incentives, workplace charity campaigns, retail-transaction solicitations, and the design of experiments on prosocial behavior. Managers can manipulate the extent to which employees are aware of and affected by each other in hopes of increasing productivity, and our results suggest externalities that should be factored into such

manipulations. Workplace charity campaigns, such as those run by the United Way, could convene meetings of groups of workers before requesting donations. Charities that solicit donations through retail transactions may want to focus their efforts in busy shopping centers. Researchers should consider the possibility that audiences have effects not only through social sanctioning but may also simply increase the degree to which subjects think about others. For all of these audiences it would be worth exploring whether other aspects of working with others have similar spillover effects on prosocial behavior.

References

- Akerlof, George A. and Rachel E. Kranton, "Economics and Identity," *Quarterly Journal of Economics*, 2000, 115, 715–753.
- Alpizar, Francisco, Fredrik Carlsson, and Olof Johansson-Stenman, "Anonymity, Reciprocity, and Conformity: Evidence from Voluntary Contributions to a National Park in Costa Rica," *Journal of Public Economics*, 2008, 92, 1047–1060.
- Andreoni, James and Douglas B. Bernheim, "Social Image and the 50–50 Norm: A Theoretical and Experimental Analysis of Audience Effects," *Econometrica*, 2009, 77, 1607–1636.
- and Justin M. Rao, "The power of asking: How communication affects selfishness, empathy, and altruism," *Journal of Public Economics*, 2011, 95(7-8), 513 – 520.
- and Ragan Petrie, "Public Goods Experiments Without Confidentiality: a Glimpse into Fund-raising," *Journal of Public Economics*, 2004, 88, 1605–23.
- Anik, Lalin, Lara B Aknin, Michael I Norton, Elizabeth W Dunn, and Jordi Quoidbach, "Prosocial bonuses increase employee satisfaction and team performance," *PloS one*, 2013, 8 (9), e75509.
- Ariely, Dan, Anat Bracha, and Stephan Meier, "Doing Good or Doing Well? Image Motivation and Monetary Incentives in Behaving Prosocially," *American Economic Review*, 2009, 99, 544–555.
- Auten, G.E., H. Sieg, and C.T. Clotfelter, "Charitable giving, income, and taxes: an analysis of panel data," *The American Economic Review*, 2002, 92 (1), 371–382.
- Bakija, Jon and Bradley T. Heim, "How Does Charitable Giving Respond To Incentives And Income? New Estimates From Panel Data," *National Tax Journal*, 2011, 64(2), 615–50.
- Bandiera, Oriana, Iwan Barankay, and Imran Rasul, "Social incentives in the workplace," *The Review of Economic Studies*, 2010, 77 (2), 417–458.
- Barraza, Jorge A and Paul J Zak, "Empathy Toward Strangers Triggers Oxytocin Release and Subsequent Generosity," *Annals of the New York Academy of Sciences*, 2009, 1167 (1), 182–189.
- , Michael E McCullough, Sheila Ahmadi, and Paul J Zak, "Oxytocin Infusion Increases Charitable Donations Regardless of Monetary Resources," *Hormones and Behavior*, 2011, 60 (2), 148–151.
- Bateson, Melissa, Daniel Nettle, and Gilbert Roberts, "Cues of Being Watched Enhance Cooperation in a Real-world Setting," *Biology Letters*, 2006, 2, 412–414.
- Bicchieri, Cristina and Erte Xiao, "Do the Right Thing: But Only if Others Do So," *Journal of Behavioral Decision Making*, 2009, 22, 191–208.
- Bochet, Olivier, Talbot Page, and Louis Putterman, "Communication and Punishment in Voluntary Contribution Experiments," *Journal of Economic Behavior & Organization*, 2006, 60, 11–26.
- Bohnet, Iris and Bruno S. Frey, "Social Distance and Other-Regarding Behavior in Dictator Games: Comment," *American Economic Review*, 1999, 89(1), 335–339.
- Borjas, George J, "Ethnic Capital and Intergenerational Mobility," *The Quarterly Journal of Economics*, 1992, 107 (1), 123–150.
- Bruhin, Adrian, Lorenz Goette, Simon Haenni, and Lingqing Jiang, "Spillovers of Prosocial Motivation: Evidence from an Intervention Study on Blood Donors," 2016.
- Card, David, Alexandre Mas, Enrico Moretti, and Emmanuel Saez, "Inequality at Work: The Effect of Peer Salaries on Job Satisfaction," *American Economic Review*, 2012, 102 (6), 2981–3003.
- Chetty, Raj, Nathaniel Hendren, Patrick Kline, and Emmanuel Saez, "Where is the Land of Opportunity? The Geography of Intergenerational Mobility in the United States," *Quarterly Journal of Economics*, 2014, 129(4), 1553–1623.

- Chiaburu, Dan S and David A Harrison**, “Do peers make the place? Conceptual synthesis and meta-analysis of coworker effects on perceptions, attitudes, OCBs, and performance.,” *Journal of Applied Psychology*, 2008, *93* (5), 1082.
- Coleman, James S**, “Social capital in the creation of human capital,” *American journal of sociology*, 1988, pp. S95–S120.
- Corgnet, Brice, Roberto Hernán Gonzalez, and Ricardo Mateo**, “Cognitive Reflection and the Diligent Worker: An Experimental Study of Millennials,” *PLoS ONE*, 2015, *10* (11), e0141243.
- DellaVigna, Stefano, James A. List, and Ulrike Malmendier**, “Testing for Altruism and Social Pressure in Charitable Giving,” *The Quarterly Journal of Economics*, 2012, *127*(1), 1–56.
- Denant-Boemont, Laurent, David Masclet, and Charles Noussair**, “Announcement, Observation, and Honesty in the Voluntary Contributions Game,” *Pacific Economic Review*, 2011, *16*, 207–228.
- Dolan, Paul and Matteo M. Galizzi**, “Like Ripples on a Pond: Behavioral Spillovers and Their Implications for Research and Policy,” *Journal of Economic Psychology*, 2015, *47* (1), 1–16.
- Drouvelis, Michalis and Benjamin M. Marx**, “Context-Dependent Donation Preferences,” mimeo.
- **and Brit Grosskopf**, “The effects of induced emotions on pro-social behaviour,” *Journal of Public Economics*, 2016, *134*, 1–8.
- **, Robert Metcalfe, and Nattavudh Powdthavee**, “Can priming cooperation increase public good contributions?,” *Theory and Decision*, 2015, *79* (3), 479–492.
- Ekström, Mathias**, “Do Watching Eyes Affect Charitable Giving? Evidence from a Field Experiment,” *Experimental Economics*, 2012, *15* (3), 530–546.
- Erkal, Nisvan, Lata Gangadharan, and Nikos Nikiforakis**, “Relative Earnings and Giving in a Real-Effort Experiment,” *The American Economic Review*, 2011, *101* (7), 3330–3348.
- Ernest-Jones, Max, Daniel Nettle, and Melissa Bateson**, “Effects of Eye Images on Everyday Cooperative Behavior: a Field Experiment,” *Evolution and Human Behavior*, 2011, *32*, 172–178.
- Falk, Armin**, “Gift exchange in the field,” *Econometrica*, 2007, *75* (5), 1501–1511.
- Frederick, Shane**, “Cognitive Reflection and Decision Making,” *Journal of Economic Perspectives*, 2005, *19*(4), 25–42.
- Garbarino, Ellen, Robert Slonim, and Carmen Wang**, “The Multidimensional Effects of a Small Gift: Evidence from a Natural Field Experiment,” *Economics Letters*, July 2013, *120* (1), 83–61.
- Gneezy, Uri, Alex Imas, and Kristóf Madarász**, “Conscience Accounting: Emotion Dynamics and Social Behavior,” *Management Science*, 2014, *60* (11), 2645–2658.
- **, Stephan Meier, and Pedro Rey-Biel**, “When and Why Incentives (Don’t) Work to Modify Behavior,” *Journal of Economic Perspectives*, 2011, *25* (4), 191–210.
- Greiner, Ben**, “Subject Pool Recruitment Procedures: Organizing Experiments with ORSEE,” *Journal of the Economic Science Association*, 2015, *1*(1), 114–25.
- Haley, Kevin J. and Daniel M. T. Fessler**, “Nobody’s Watching? Subtle Cues Affect Generosity in an Anonymous Economic Game,” *Evolution and Human Behavior*, 2005, *26*, 245–256.
- Heap, Shaun P. Hargreaves and Daniel John Zizzo**, “The Value of Groups,” *The American Economic Review*, 2009, *99* (1), 295–323.
- Huck, Steffen and Imran Rasul**, “Matched fundraising: Evidence from a natural field experiment,” *Journal of Public Economics*, 2011, *95*, 351–362.
- **, —, and Andrew Shephard**, “Comparing charitable fundraising schemes: Evidence from a field experiment and a structural model,” *American Economic Journal: Economic Policy*, 2015, *7* (2), 326–69.

- IJzendoorn, Marinus H Van, Renske Huffmeijer, L Ra Alink, Marian J Bakermans-Kranenburg, and Mattie Tops**, “The impact of oxytocin administration on charitable donating is moderated by experiences of parental love-withdrawal,” *Front. Psychol*, 2011, *2* (258), 10–3389.
- Karlan, Dean and John A. List**, “Does Price Matter in Charitable Giving? Evidence from a Large-Scale Natural Field Experiment,” *American Economic Review*, 2007, *97* (5), 1774–1793.
- Kessler, Judd B**, “Announcements of Support and Public Good Provision,” *American Economic Review*, forthcoming.
- Knack, Stephen and Philip Keefer**, “Does Social Capital Have an Economic Payoff? A Cross-Country Investigation,” *The Quarterly Journal of Economics*, 1997, pp. 1251–1288.
- Landry, Craig E., Andreas Lange, John A. List, Michael K. Price, and Nicholas G. Rupp**, “Toward an Understanding of the Economics of Charity: Evidence from a Field Experiment,” *The Quarterly Journal of Economics*, 2006, *121*(2), 747–782.
- , —, —, —, —, and —, “Is a Donor in Hand Better Than Two in the Bush? Evidence from a Natural Field Experiment,” *American Economic Review*, 2010, *100*(3), 958–83.
- List, John A and David Lucking-Reiley**, “The effects of seed money and refunds on charitable giving: Experimental evidence from a university capital campaign,” *Journal of Political Economy*, 2002, *110* (1), 215–233.
- Mas, Alexandre and Enrico Moretti**, “Peers at Work,” *American Economic Review*, 2009, *99* (1), 112–45.
- McCannon, Bryan and Zachary Rodriguez**, “The Impact of Micro-Finance on Pro-Social Behaviors: Experimental Evidence of Public Goods Contributions in Uganda,” *mimeo*, *West Virginia University*, 2016.
- Meer, Jonathan**, “Brother, Can You Spare a Dime? Peer Pressure in Charitable Solicitation,” *Journal of Public Economics*, 2011, *95* (7-8), 926–941.
- and **Harvey Rosen**, “The ABCs of Charitable Solicitation,” *Journal of Public Economics*, 2011, *95* (5-6), 363–371.
- Niederle, Muriel and Lise Vesterlund**, “Explaining the Gender Gap in Math Test Scores: The Role of Competition,” *Journal of Economic Perspectives*, 2010, *24*(2), 129–44.
- Pan, Xiaofei Sophia and Daniel Houser**, “Cooperation During Cultural Group Formation Promotes Trust Towards Members of Out-Groups,” *Proceedings of the Royal Society B*, 2013, *280* (20130606).
- Putnam, Robert**, *Bowling alone: the collapse and revival of American community.*, Simon and Schuster, New York, 2000.
- Rege, Mari and Kjetil Telle**, “The Impact of Social Approval and Framing on Cooperation in Public Good Situations,” *Journal of Public Economics*, 2004, *88*, 1625–44.
- Samek, Anya Savikhin and Roman M Sheremeta**, “Recognizing Contributors: An Experiment on Public Goods,” *Experimental Economics*, 2014, *17* (4), 673–690.
- Scharlemann, Jörn PW, Catherine C Eckel, Alex Kacelnik, and Rick K Wilson**, “The value of a smile: Game theory with a human face,” *Journal of Economic Psychology*, 2001, *22* (5), 617–640.
- Shang, Jen and Rachel Croson**, “A Field Experiment in Charitable Contribution: the Impact of Social Information on the Voluntary Provision of Public Goods,” *Economic Journal*, 2009, *119*, 1422–1439.
- Shariff, Azim and Ara Noranzayan**, “God is Watching You: Priming God Concepts Increases Prosocial Behavior in an Anonymous Dictator Game,” *Psychological Science*, 2007, *18*(9), 803–9.
- Soetevent, Adriaan R.**, “Anonymity in Giving in a Natural Context – a Field Experiment in 30 Churches,” *Journal of Public Economics*, 2005, *89*, 2301–2323.

- Tajfel, H and JC Turner**, “The Social Identity Theory of Intergroup Behavior.,” *Psychology of Intergroup Relations*, 1986.
- Tajfel, Henri**, “Experiments in Intergroup Discrimination,” *Scientific American*, 1970, *223* (5), 96–102.
- , **Michael G Billig, Robert P Bundy, and Claude Flament**, “Social Categorization and Intergroup Behaviour,” *European journal of social psychology*, 1971, *1* (2), 149–178.
- Tonin, Mirco and Michael Vlassopoulos**, “Sharing one’s fortune? An experimental study on earned income and giving,” *Journal of Behavioral and Experimental Economics*, 2017, (66), 112–118.
- Wozniak, David, William T. Harbaugh, and Ulrich Mayr**, “The Menstrual Cycle and Performance Feedback Alter Gender Differences in Competitive Choices,” *Journal of Labor Economics*, 2014, *32*(1), 161–98.
- Zak, Paul J, Angela A Stanton, and Sheila Ahmadi**, “Oxytocin Increases Generosity in Humans,” *PLoS one*, 2007, *2* (11), e1128.

Appendix A: Experiment Instructions

[Note: These are the written instructions as presented to subjects facing the “Individual, Private” treatment in Sequence 1. Amendments to the remaining treatments are given in square brackets. For Sequence 2, the order of the tasks was as follows: hard math, easy math, hard word, easy word].

INSTRUCTIONS

Welcome! You are about to take part in an experiment. This experiment is run by the “Birmingham Experimental Economics Laboratory” and has been financed by various research foundations. Just for showing up you have already earned £2.50. You can earn additional money depending on the decisions made by you and other participants. It is therefore very important that you read these instructions with care.

It is important that you remain silent and do not look at other people’s work. If you have any questions, or need assistance of any kind, please raise your hand and an experimenter will come to you. You may use the provided scrap paper but no phones, calculators, or other devices. If you use a device, talk, laugh, exclaim out loud, etc., you will be asked to leave and you will not be paid. We expect and appreciate your following of these rules.

We will first jointly go over the instructions. After we have read the instructions, you will have time to ask clarifying questions. Please do not touch the computer or its mouse until you are instructed to do so. Thank you.

This experiment consists of four different timed tasks. You will be paid a fixed amount of money for each correct answer you provide in each task. [**Group:** At the beginning of the experiment, you will be put in a group with two other people selected at random. You will be paid a fixed amount of money for each correct answer your group of 3 people provides in each task. Your group earnings will be divided equally among the 3 group members.] The total amount of money you will earn from this experiment will be £2.50 for showing up plus the sum of your earnings from each task of the experiment.

After Task 4 you will be told how many correct responses you gave in each of the tasks. [**Public:** You will then be asked by the experimenter to stand up and tell the others in the room the number of correct answers you provided in each of the four tasks.] After this you will collect your earnings.

Following these instructions you will find the instructions for Task 1 of the experiment. You will receive new instructions for the other tasks once everyone in the room has completed Task 1.

Task 1 (These instructions also appeared onscreen)

Task 1 consists of solving 2-number multiplication problems like the following example:

$$5 \times 7 = 35. \qquad 9 \times 8 = 72.$$

You will have 5 minutes to provide answers.

You will be paid 3 pence for each correct answer provided during the 5 minute time limit.

[Group: Your group will be paid 3 pence for each correct answer provided during the 5 minute time limit. These earnings will be divided equally among the 3 members of your group. In other words, you will receive 1 penny for each correct answer provided by you or either of your group members.]

To answer a problem, you will simply type the numbers on the keyboard, then press OK and another problem will appear. You can choose not to answer a question by pressing the OK button. The answer will then be recorded as being incorrect and you will be moved to the next problem. To help with time management, there will be a clock counting down the seconds for the 5 minute duration.

Task 2 (These instructions also appeared onscreen)

Task 2 consists of solving 2-number multiplication problems like the following example:

$$10 \times 97 = 970. \qquad 20 \times 30 = 600.$$

You will have 5 minutes to provide answers.

You will be paid 21 pence for each correct answer provided during the 5 minute time limit.

[Group: Your group will be paid 21 pence for each correct answer provided during the 5 minute time limit. These earnings will be divided equally among the 3 members of your group. In other words, you will receive 7 pence for each correct answer provided by you or either of your group members.]

To answer a problem, you will simply type the numbers on the keyboard, then press OK and another problem will appear. You can choose not to answer a question by pressing the OK button. The answer will then be recorded as being incorrect and you will be moved to the next problem. To help with time management, there will be a clock counting down the seconds for the 5 minute duration.

Task 3 (These instructions also appeared onscreen)

Task 3 consists of arranging pairs of letters to form words like the following examples:

PO, TS = POTS. GE, MA = MAGE.

You must use all the letters. You can change the order of the pairs but you cannot change the order of the two letters within each pair. You will have 5 minutes to provide answers.

You will be paid 3 pence for each correct answer provided during the 5 minute time limit.

[Group: Your group will be paid 3 pence for each correct answer provided during the 5 minute time limit. These earnings will be divided equally among the 3 members of your group. In other words, you will receive 1 penny for each correct answer provided by you or either of your group members.]

To answer a problem, you will simply type the word on the keyboard, then press OK and another problem will appear. You can choose not to answer a question by pressing the OK button. The answer will then be recorded as being incorrect and you will be moved to the next problem. To help with time management, there will be a clock counting down the seconds for the 5 minute duration.

Task 4 (These instructions also appeared onscreen)

Task 4 consists of arranging pairs of letters to form words like the following examples:

TR, EA, TS, RE = RETREATS. CU, FF, LI, NK = CUFFLINK.

You must use all the letters. You can change the order of the pairs but you cannot change the order of the two letters within each pair. You will have 5 minutes to provide answers.

You will be paid 21 pence for each correct answer provided during the 5 minute time limit.

[Group: Your group will be paid 21 pence for each correct answer provided during the 5 minute time limit. These earnings will be divided equally among the 3 members of your group. In other words, you will receive 7 pence for each correct answer provided by you or either of your group members.]

To answer a problem, you will simply type the word on the keyboard, then press OK and another problem will appear. You can choose not to answer a question by pressing the OK button. The answer will then be recorded as being incorrect and you will be moved to the next problem. To help with time management, there will be a clock counting down the seconds for the 5 minute duration.

At the end of Task 4, the experimenter makes the following announcement: You can now see the number of correct answers you gave in each of the four tasks. Please give me a moment to print the results.

[Public: I will now call you one by one to stand up and tell the others in the room the number of correct answers you provided in each of the four tasks.]

OK, we are almost done! You will now be given an opportunity to donate some of your income from the experiment to a charity, and last, you will be asked to complete a survey.

After this announcement, subjects received the following instructions onscreen:

Thank you, you have completed all 4 Tasks. Your total earnings from today's experiment (including your £2.50 show-up fee) sum to £[*Autofill*].

Would you like to donate some of your earnings to Acorns Children's Hospice of Birmingham [**Written with hyperlink to separate linked screen**]? If so, please enter the amount (between £0 and £[*Autofill*]) in the box provided.



[Input field]

[OK button]

Thank you. The experimenter will now ask you to complete a short questionnaire and will give you the payment for your earnings.

Separate linked screen



Acorns provides specialist care for babies, children and young people who are life-limited or life-threatened. As well as this, Acorns offers support for the whole family; including mums, dads, brothers and sisters, helping them to cope with the huge challenges they face living with and caring for a child with complex disabilities. In order to continue its work, the charity must raise more than £9 million every year, and it relies on the public for the vast majority of that amount.

[OK button]

Appendix B: Additional Figures and Tables

Table B.1: Covariates Balanced Across Treatments

	(1)	(2)	(3)	(4)	(5)	(6)
	CRTscore	Male	Age	UK	Married	Knows Labmate
Group, Private	0.079 (0.171)	0.085 (0.084)	-0.304 (0.816)	0.110 (0.083)	0.040 (0.045)	0.034 (0.077)
Individual, Public	0.209 (0.187)	0.233*** (0.082)	0.800 (0.799)	-0.007 (0.085)	0.071 (0.049)	0.024 (0.076)
Group, Public	0.118 (0.191)	0.093 (0.085)	-0.487 (0.752)	0.124 (0.083)	0.001 (0.040)	-0.040 (0.074)
Constant	1.143*** (0.128)	0.443*** (0.060)	21.957*** (0.582)	0.529*** (0.060)	0.057** (0.028)	0.271*** (0.054)
N	282	282	280	282	281	282
Adj. R-Squared	-0.006	0.018	0.001	0.004	0.001	-0.007

Heteroskedasticity-robust standard errors are in parentheses. The “outcomes” are, respectively, the subject’s Cognitive-Reflexive Test score, a dummy for male, age, a dummy for United Kingdom nationality, a dummy for being married, and a dummy for knowing another subject in the laboratory.

Table B.2: Characteristics by Choice of Whether to Donate

	(1)	(2)	(3)	(4)	(5)	(6)
	CRTscore	Male	Age	UK	Married	Knows Labmate
Any donation	-0.082 (0.135)	0.095 (0.062)	1.038* (0.528)	-0.107* (0.060)	0.072** (0.031)	-0.001 (0.056)
Constant	1.297*** (0.108)	0.485*** (0.050)	21.297*** (0.376)	0.653*** (0.048)	0.040** (0.019)	0.277*** (0.045)
N	282	282	280	282	281	282
Adj. R-Squared	-0.002	0.005	0.008	0.007	0.012	-0.004

Heteroskedasticity-robust standard errors are in parentheses. The “outcomes” are, respectively, the subject’s Cognitive-Reflexive Test score, a dummy for male, age, a dummy for United Kingdom nationality, a dummy for being married, and a dummy for knowing another subject in the laboratory. “Any donation” is a dummy variable, and its coefficient captures the difference from the average value of the characteristic among subjects choosing to not donate.

Table B.3: Performance by Type of Task
Items Answered Correctly

	(1)	(2)	(3)	(4)	(5)	(6)	(7)
	Own Earnings	Easy Math	Easy Words	Easy Tasks	Hard Math	Hard Words	Hard Tasks
Group, Private	0.770 (0.751)	0.674 (8.097)	4.887 (3.357)	5.560 (9.800)	0.010 (1.359)	2.864* (1.583)	2.874 (2.394)
Individual, Public	0.828 (0.544)	4.644 (7.416)	4.957** (2.256)	9.601 (8.494)	0.597 (0.906)	1.973 (1.235)	2.570 (1.817)
Group, Public	-0.232 (0.539)	-5.460 (7.792)	-0.607 (3.591)	-6.068 (9.406)	-1.425 (1.115)	1.187 (1.616)	-0.239 (1.673)
Constant	12.841*** (0.414)	150.243*** (6.164)	99.086*** (2.433)	249.329*** (7.260)	8.643*** (0.919)	16.886*** (0.788)	25.529*** (1.130)
N	282	282	282	282	282	282	282
Adj. R-Squared	0.007	-0.004	0.011	0.002	0.007	0.006	0.007

Items Attempted

	(1)	(2)	(3)	(4)	(5)	(6)
	Easy Math	Easy Words	Easy Tasks	Hard Math	Hard Words	Hard Tasks
Group, Private	1.551 (7.841)	4.705 (3.199)	6.256 (9.550)	11.520*** (3.745)	5.896* (2.991)	17.416*** (5.516)
Individual, Public	6.237 (7.292)	5.716** (2.394)	11.954 (8.560)	3.461* (1.699)	4.968** (1.817)	8.429** (3.002)
Group, Public	-4.867 (7.815)	-1.158 (3.433)	-6.025 (9.561)	-0.112 (1.337)	0.685 (2.519)	0.573 (3.388)
Constant	153.157*** (6.129)	102.129*** (2.287)	255.286*** (7.306)	12.271*** (0.662)	26.243*** (1.333)	38.514*** (1.323)
N	282	282	282	282	282	282
Adj. R-Squared	-0.003	0.017	0.005	0.042	0.023	0.053

Items Left Blank

	(1)	(2)	(3)	(4)	(5)	(6)
	Easy Math	Easy Words	Easy Tasks	Hard Math	Hard Words	Hard Tasks
Group, Private	0.262 (0.334)	-0.100 (0.171)	0.162 (0.366)	11.863*** (3.602)	2.790 (2.172)	14.654*** (4.633)
Individual, Public	0.569 (0.417)	0.090 (0.193)	0.659 (0.418)	2.909** (1.034)	2.781* (1.324)	5.691*** (1.907)
Group, Public	-0.151 (0.273)	-0.353** (0.154)	-0.505 (0.294)	1.586 (1.153)	0.095 (1.614)	1.681 (2.399)
Constant	0.571** (0.246)	0.614*** (0.180)	1.186*** (0.286)	0.429** (0.176)	6.543*** (1.134)	6.971*** (1.175)
N	282	282	282	282	282	282
Adj. R-Squared	0.011	0.006	0.022	0.047	0.007	0.046

Standard errors clustered by subject are in parentheses. In Models (1) and (2), the dependent variables are the number of correct answers in the easy and hard tasks, respectively. In Models (3)-(6), we include as dependent variables the number of correct responses in each of the four tasks. This allows us to disentangle which task can explain better any potential differences recorded in the pooled data reported in the first two models.

Table B.4: Effects on Quintiles of Total Earnings

Percentile:	20th	40th	60th	80th
Group, Private	0.580 (0.594)	0.300 (0.623)	0.570 (0.456)	0.700 (0.815)
Individual, Public	-0.240 (0.596)	0.630 (0.625)	1.470*** (0.458)	1.080 (0.818)
Group, Public	0.430 (0.600)	0.350 (0.630)	-0.420 (0.461)	-0.680 (0.823)
Constant	10.620*** (0.423)	12.150*** (0.444)	13.470*** (0.325)	15.090*** (0.580)
N	282	282	282	282
Adj. R-Squared				

Notes: In all models the dependent variable is an individual's total earnings. Each column displays results of a quantile regression for the quantile listed at the top of the column.

Table B.5: Donation Effects by Cognitive Level

	All Subjects	CRT=0	CRT=1	CRT=2	CRT=3
	(1)	(2)	(3)	(4)	(5)
Group, Private	0.810** (0.369)	1.347* (0.801)	0.417 (0.537)	0.342 (0.602)	1.581 (0.961)
Individual, Public	1.110** (0.535)	2.231* (1.304)	0.653 (0.657)	0.070 (0.708)	1.601 (1.347)
Group, Public	0.649* (0.367)	1.279* (0.751)	0.205 (0.522)	-0.171 (0.764)	1.489 (0.961)
Constant	-0.296 (0.321)	-0.823 (0.722)	0.372 (0.404)	0.034 (0.450)	-1.025 (0.937)
N	282	93	75	66	48

Heteroskedasticity-robust standard errors are in parentheses. All models are Tobit models for donations, and reported estimates are marginal effects on observed donations. Each model reports estimates for a group with a different CRT score.

Table B.6: Donation Invariance to Own and Others' Earnings

	All Group		Group, Private		Group, Public	
	(1) Donation	(2) Any Donation	(3) Donation	(4) Any Donation	(5) Donation	(6) Any Donation
Groupmembers' Earnings	-0.005 (0.210)	-0.109 (0.109)	-0.239 (0.337)	-0.259 (0.182)	-0.088 (0.331)	0.017 (0.159)
Own Earnings	-0.040 (0.131)	-0.076 (0.071)	-0.157 (0.226)	-0.188 (0.115)	-0.145 (0.191)	0.030 (0.101)
Own Earnings * Groupmembers'	0.003 (0.016)	0.009 (0.008)	0.014 (0.025)	0.021 (0.013)	0.018 (0.027)	-0.003 (0.013)
Constant	1.138 (1.747)	1.644* (0.962)	3.543 (3.124)	3.092* (1.663)	1.510 (2.394)	0.469 (1.276)
N	141	141	72	72	69	69
Adj. R-Squared	-0.017	-0.011	-0.027	0.012	-0.004	-0.038

Heteroskedasticity-robust standard errors are in parentheses.

Table B.7: Spillover effects on donations – Controlling for earnings

	(1)	(2)	(3)	(4)
	Any Donation	Amount	Cond. Amount	Tobit
Group, Private	0.207** (0.081)	0.324 (0.203)	0.080 (0.296)	0.815** (0.363)
Individual, Public	0.105 (0.084)	0.764** (0.352)	1.014* (0.534)	1.116** (0.526)
Group, Public	0.196** (0.082)	0.192 (0.209)	-0.087 (0.318)	0.648* (0.371)
Constant	0.504*** (0.153)	0.760 (0.606)	1.473 (0.980)	-0.205 (0.910)
N	282	282	181	282
Adj. R-Squared	0.016	0.012	0.028	

Heteroskedasticity-robust standard errors are in parentheses. Columns (1)-(3) report results of OLS regressions. Dependent variables, by model, are an indicator for a nonzero donation (1), the amount donated (2), and the amount donated conditional on a nonzero donation (3). Model (4) is a Tobit model for donations, and reported coefficients are marginal effects on observed donations. All regressions the same as in Table 5 except for the inclusion of a control variable for the subjects' total payoff.

Table B.8: Income Patterns by Whether Subject Knows Labmates

	All Subjects	Subject Doesn't Know Others	Subject Knows Others
	(1)	(2)	(3)
Group, Private	0.770 (0.573)	0.717 (0.674)	0.801 (1.073)
Individual, Public	0.828 (0.538)	0.722 (0.647)	1.012 (0.946)
Group, Public	-0.232 (0.553)	-0.034 (0.648)	-0.750 (1.095)
Constant	12.841*** (0.349)	12.622*** (0.430)	13.429*** (0.563)
N	282	204	78
Adj. R-Squared	0.007	-0.003	-0.003

Heteroskedasticity-robust standard errors are in parentheses.