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**by**

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# Short- and Long-run Effects of External Interventions on Trust.\*

Igor Asanov<sup>†</sup> and Simone Vannuccini<sup>‡</sup>

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## Abstract

We experimentally analyze the effects of external interventions such as subsidy and targeting on investment decisions during the intervention and after. We employ a multi-period version of the trust (investment) game (Berg et al., 1995) introducing either the monetary incentives for contribution or providing a suggestion about the level of investment. The results of the experiment indicate that targeting is an effective instrument to promote trustful behavior, whereas subsidy policy is not effective both in short- and long-run. Therefore we suggest considering a targeting policy as one of the instruments that can foster trustful behavior.

*Keywords:* Trust Game; Experiment; Policy; Subsidy; Academic Spin-offs

*JEL Classification:* C92, L50, D80

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# 1 Introduction

In 1998, Stanford University licenses the *PageRank* patent to one of its newly established spin-off companies. This investment initiates the growth of one of the worlds' largest high tech company *Google* that soon revolutionizes the IT market and changes the economy. Besides public economic impact, this investment brings private financial benefits to Stanford that in large extent include voluntary financing of research scholarships and common projects.<sup>1</sup>

The success of Google explains why governments often intervene aiming to foster the academic spin-off creation and knowledge commercialization. Typically, such intervention takes the form of subsidy-policy that comprises two phases: First, a university receives a subsidy if it invests in the spin-off; Secondly, the successful spin-off gains additional finances from the government.<sup>2</sup>

Alternative forms of policy such as targeting are rarely considered, though they may not involve subsidy spending. Moreover, since the policy makers are often focused on immediate consequences of the interventions, the long-term, post-intervention potential costs are not taken into account. We attempt to fill this gap using controlled laboratory experiment that allows to make a direct comparison of different policies' efficiency in the short- and in the long-run.

In this experiment we analyze the effects of external intervention such as subsidy and targeting on the investment decision during the intervention and after. We employ a multi-period version of the trust (investment) game (Berg et al., 1995) introducing either the monetary incentives for contribution or providing a suggestion about the level of investment. The experiment consists of three blocks with policy intervention in the second one that let us to consider immediate as well as post-intervention effects.

The study offers three main original contributions: First, we analyze the effect of non-monetary intervention in form of suggestion on trustful behavior; Second, we compare the effect of non-monetary policy to monetary ones; Third, we provide an analysis of the long-run effects of external interventions on trust.

More specifically, we aim to answer the next four questions: (1) Do

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<sup>1</sup>For instance, in 2008, Google paid approximately \$1,881,400 to Stanford University out of which only \$426,950 payments related to the license of patents. The largest part of the payments - about \$1,246,000 - was donations for scholarships and other philanthropic endeavors (Wikinvest.com, 2009).

<sup>2</sup>See, for example, programs such as "Small Business Technology Transfer" (SBTT) in the United States and "Existenzgründungen aus der Wissenschaft" (EXIST) in Germany.

non-monetary intervention such as suggestion increase investment activity during and after they are introduced? (2) Is subsidy policy an efficient mean to foster investment activity in the short-run? (3) Is a low level of investment required to receive a subsidy detrimental for an investment? (4) Does the subsidy policy have a negative impact on investment level after the policy termination?

We find that non-monetary policy in form of suggestion increase investment activity during the intervention and we do not find any detrimental effects afterwards. Subsidy policy, however, does not significantly affect the level of trust in the short- or in the long-run. We associate the ineffectiveness of subsidy policy with two regularities: Subjects show low propensity to follow this policy and if subjects follow it, they send mostly the lowest amount required to get the subsidy.

We also find indirect evidence that the monetary policy is ineffective not because of the presence of the subsidy itself but rather from the fact that monetary reward is conditioned on a certain behavior: Subjects that unconditionally receive subsidies do not show a significantly different level of trustworthiness. We conclude that targeting policy should be considered as an effective tool to foster investment activity.

The rest of the paper proceeds as follows: Section two provides a short review of the relevant literature. Section three describes the theoretical framework and the hypotheses. Section four presents the experimental design. Section five provides the results of the experiment. Section six then discusses the findings, followed by some final remarks.

## **2 Related Literature**

The paper builds on three different strands of literature.

First, it relates to studies on interaction between intrinsic and extrinsic motivation. From the early research of Titmuss (1970) on blood donations to the experiment of Andreoni (1993) on public good provision, the studies point out the potential detrimental effects of external interventions on intrinsic motivation. For instance, in a meta-analysis of experimental studies on external incentives and intrinsic motivation, Deci et al. (1999) indicate the presence of negative effects that are particularly relevant in case of tangible rewards.

Bowles and Polania-Reyes (2012), however, come to a different conclusion evaluating the results of experiments on the relation between incentives and social preferences. They note that the effect of the incentives depends on the pre-existing social framework and can be both negative and posi-

tive. Gneezy et al. (2011) extend this discussion urging to consider both the potential long-term costs and benefits of external interventions.

The second strand of literature looks at the role that trust plays in investment decisions. Trust is involved in almost every economic transaction (Arrow, 1972) and, indeed, the empirical evidence suggests that the trust is crucial for venture capital investments (Bottazzi et al., 2011), mutual investment decisions (Felli et al., 2010) and has a positive association with the level of investment across countries (Knack and Keefer, 1997).

The trust (investment) game that we employ in the experiment mirrors the investment situation with imperfect contracts. The behavior in this game varies across countries with different economic characteristics (Johnson and Mislin, 2011). Moreover, the trustful behavior in this game correlates with the differences in investment propensity between countries – for instance, Germany and France (Willinger et al., 2003) or Gulf region and Western countries (Bohnet et al., 2010) – that make it possible to better understand the variation in the investment rates across nations.

Thirdly, this paper is closely related to the studies of the interaction between external incentives and trustful behavior. Fehr and List (2004) find that the threat to punish increases trustworthiness, but the punishment crowds out trustworthy behavior. Furthermore, Bohnet et al. (2001) find that the threat of potential contract enforcement crowds in trustworthiness, although this effect depends on the level of enforcement.

Studying the effect of various incentives on trustful behavior Char-ness et al. (2008) allow a third-party not only to punish but to reward as well. The experimental results corroborate the hypothesis that the threat of punishment increases trust and trustworthiness. However, the effect of reward on trust is rather ambiguous.

Gächter et al. (2011) further extend the research on the effect of punishment and rewards on trustworthiness. Using a multi-phase gift-exchange game they find that trustworthiness (exerted effort) increases both in the presence of a fine or a bonus. Nevertheless, the effect of the bonus is much smaller than the effect of the fine due to the crowding out: Under the bonus condition subjects tend to choose an effort not higher than a best-reply level.

As concerns the effect of non-monetary incentives on trust, Bracht and Feltovich (2009) show that the information about the previous actions of others can enhance cooperation.<sup>3</sup> Moreover, Berg et al. (1995) provide evidence that even an aggregated information about previous behavior – information about the average amount sent by other subjects –

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<sup>3</sup>In addition, Duffy and Feltovich (2010) find that the recommendation by third-party affects subjects behavior in the two-player game of Chicken.

can strengthen trustful relations. Similarly, Thöni and Gächter (2012) show that peer-effects have a significant influence on the trust level and suggest conformism as an explanation of this phenomenon.

### 3 Theory and Implications

#### 3.1 The Game

We use a version of the trust (investment) game. In the original trust game (Berg et al., 1995), two players interact with each other: player 1 (the trustor) decides which amount of his initial endowment  $E$  to send (to give) to player 2 (the trustee). The amount sent  $s$  is multiplied by a certain factor  $m$  and player 2 receives the multiplied sum. Player 2, in turn, chooses how much to return  $R$  of the amount received. See Figure 1 for the structure of the game and a description of the payoffs  $\pi$  of players 1 and 2.

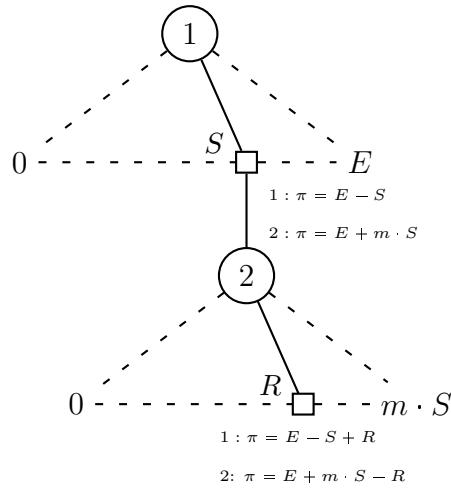


Figure 1: Trust (investment) game.

In our version of the investment game, an external intervention is introduced. This intervention is devised alternatively as either a subsidy or a suggestion. The subsidy  $Z$  is obtained by both players if the contribution of player 1 is greater than or equal to a certain threshold  $T$  (figure 2 describes this version of the game). In the case of suggestion, no subsidy is available but it is suggested to send not less than a threshold level.

The game is played for several periods and consists of three blocks. Blocks 1 and 3 consist of repetitions of the standard trust game, while in the block 2 the interventions are introduced.

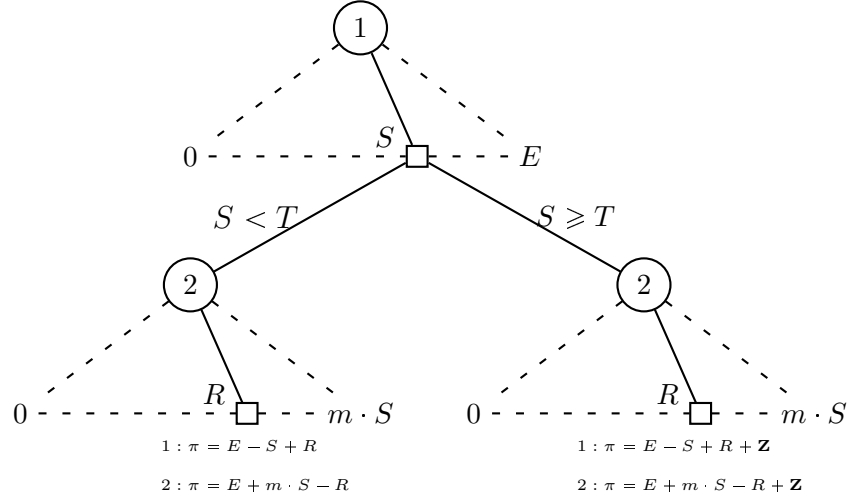


Figure 2: Trust (investment) game with subsidies.

In what follows, we outline a simple model to develop the theoretical predictions and hypotheses.

### 3.2 Trust under External Incentives

To derive the theoretical predictions, we apply backward induction solving the model from the second stage. We denote by  $v$  the value that trustor expects to receive back in the second stage of the game. This value is a function of the amount sent  $s$ . Thus, the utility function of the trustor takes the form:

$$u = E - c(s) + v(s) + o(s) + I, \quad (1)$$

where  $E$  is the player's endowment,  $c$  is the individual's cost of sending an amount  $s$ ,  $o$  is the trustor's other-regarding preferences that depend on  $s$ ,  $I$  is the effect of external incentives that can take the form of either a subsidy or a suggestion.

Let's begin the analysis with the subsidy policy. The subsidy policy is characterized by a tuple of parameters  $(Z, T)$ , indicating the size of the subsidy and the threshold (minimal) amount that the player must send to obtain this subsidy, respectively.

The subsidy offsets the costs of sending but can affect other-regarding preferences as well. We assume that the other-regarding preferences are affected by a measure  $\lambda < 0$ .<sup>4</sup> Thus, the utility function in the presence of

<sup>4</sup>We make this assumption in line with previous experimental results. See Bowles

a subsidy policy is

$$u = E - c(s) + v(s) + o(s) + 1_{\{s \geq T\}}[Z + \lambda o(s)], \quad (2)$$

where the indicator  $1_{\{s \geq T\}} = 1$  if  $s \geq T$  and zero otherwise.

The players maximize their utility so that the marginal costs of sending are equal to the marginal benefits (the values are expressed in discrete terms to account for the discontinuity in  $s = T$ ):

$$\frac{\Delta c(s)}{\Delta s} = \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{\Delta 1_{\{s \geq T\}}[Z + \lambda o(s)]}{\Delta s}, \quad (3)$$

To analyze the effect of a subsidy policy, we compare it to the case where there are no incentives. The subsidy is contingent on the relation between threshold and amount sent. We therefore consider two states (1) when the amount to be sent without incentives  $s_0$  is lower than the threshold and (2) when it is higher. We then obtain the following two relations:

$$\frac{\Delta c(s^*)}{\Delta s} = \begin{cases} \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{Z}{\Delta s} + \frac{\lambda \Delta o(s)}{\Delta s} & \text{if } s_0 < T \\ \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{\lambda \Delta o(s)}{\Delta s} & \text{if } s_0 \geq T \end{cases} \quad (4)$$

One can easily see that it is beneficial to send more whenever the amount to be sent without incentives  $s_0$  is lower than the threshold  $T$  and the direct effect of the subsidy  $\frac{Z}{\Delta s}$  is larger than the crowding out effect of the subsidy  $\frac{\lambda \Delta o(s)}{\Delta s}$ . However, if  $s_0 < T$ , there is no direct subsidy effect (the subsidy is independent from additional sending,  $\frac{Z}{\Delta s} = 0$ ), whereas the negative effect of the subsidy on other-regarding preferences is still present,  $\frac{\lambda \Delta o(s)}{\Delta s} < 0$ . We are therefore able to formulate the following two hypotheses:

**H 1.** *The amount sent is higher under external monetary incentives than without them if (1) the threshold level is higher than the amount sent in case without the incentives  $s_0 < T$  and (2) the direct effect of the subsidy is larger than the crowding out effect  $\frac{Z}{\Delta s} + \frac{\lambda \Delta o(s)}{\Delta s} > 0$ .*

**H 2.** *The amount sent is lower under external monetary incentives than without them if the threshold level is higher than the amount sent in case without the incentives  $s_0 < T$ .*

Concerning the targeting policy (suggestion), this policy is also characterized by a threshold level  $T$  (the suggested minimal level to be sent). The policy does not use subsidy but players can get an utility complying and Polania-Reyes (2012) for a discussion.



with authority (Karakostas and Zizzo, 2012).<sup>5</sup> We denote this utility by  $A$  (that is independent from  $s$ ). Thus, the senders' utility is

$$u = E - c(s) + v(s) + o(s) + 1_{\{s \geq T\}}(A), \quad (5)$$

Analyzing the players' utility function in the case of targeting policy in the same way as in 3 and 4, we obtain the next relations:

$$\frac{\Delta c(s^*)}{\Delta s} = \begin{cases} \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{A}{\Delta s} & \text{if } s_0 < T \\ \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} & \text{if } s_0 \geq T \end{cases} \quad (6)$$

If the amount sent in case without the incentives is lower than the threshold  $s_0 < T$  the players benefit by complying with authority. Therefore, they can sacrifice part of their endowment to follow the suggestion. Nevertheless, they do not benefit when  $s_0 > T$  since the utility is independent from the amount sent.

**H 3.** *The amount sent is higher under external non-monetary incentives than without them if the threshold level is higher than the amount sent in case without the incentives  $s_0 < T$ .*

Considering the long-run (post-intervention) effect of incentives, we assume that preferences are endogenous (Bowles, 1998), meaning that the preferences learned under certain circumstances stay present afterwards. Given this, we can derive from 4 the following relations for the period after the subsidy policy:

$$\frac{\Delta c(s^*)}{\Delta s} = \begin{cases} \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{\lambda \Delta o(s)}{\Delta s} & \text{if } s_0 < T \\ \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{\lambda \Delta o(s)}{\Delta s} & \text{if } s_0 \geq T \end{cases} \quad (7)$$

There is no direct effect of the subsidy  $Z$  since the subsidy policy is absent now. However, other-regarding preferences are still negatively affected  $\frac{\lambda \Delta o(s)}{\Delta s} < 0$ . Thus, we formulate:

**H 4.** *The amount sent is lower after experiencing external monetary incentives than without them.*

In a similar vein, we derive from 5 the next relations for the period after the targeting policy:

$$\frac{\Delta c(s^*)}{\Delta s} = \begin{cases} \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} + \frac{A}{\Delta s} & \text{if } s_0 < T \\ \frac{\Delta v(s)}{\Delta s} + \frac{\Delta o(s)}{\Delta s} & \text{if } s_0 \geq T \end{cases} \quad (8)$$

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<sup>5</sup>In Karakostas and Zizzo (2012), the information communicated by a third-party affects the behavior of subjects. They attribute this effect to compliance to authority. We suppose that the suggestion have a similar effect.

When the threshold level is higher than the amount sent in case without the incentives  $s_0 < T$ , the players send more after the targeting policy since they continue to gain utility complying to the authority  $\frac{A}{\Delta s} > 0$ .

**H 5.** *The amount sent is higher after experiencing external non-monetary incentives than without them if the threshold level is higher than the amount sent in case without the incentives  $s_0 < T$ .*

### 3.3 Trustworthiness under External Incentives

We represent the utility function of the trustee in the following way:

$$u = 1 - c(r) + o(r) + I, \quad (9)$$

where  $c(r)$  is the trustee's cost of returning the ratio  $r = \frac{R}{m \cdot s}$ ,  $o$  is the other regarding preferences that changes with  $r$ <sup>6</sup>,  $I$  is the effect of external intervention (subsidy or suggestion).

We assume that trustees maximize their utility. Since external intervention depends on the behavior of trustor but not on trustee's choice we obtain the following relation:

$$\frac{\partial c(r^*)}{\partial r} = \frac{\partial o(r)}{\partial r}, \quad (10)$$

We know from previous studies (Johnson and Mislin, 2011) that  $\frac{\partial c(r^*)}{\partial r \partial s} = \frac{\partial o(r)}{\partial r \partial s} > 0$ . Therefore, we can formulate the following hypothesis:

**H 6.** *The trustworthiness rate  $r$  is not different during and after the external intervention as compared to the case without it when conditioned on the amount sent by the trustor  $s$ .*

## 4 Experimental Design

The experiment was conducted at the laboratory of the Max Planck Institute of Economics in Jena (Germany) in April 2013. Seven sessions were run, each of them lasting about 60 minutes and employing 32 experimental subjects. Experimental subjects were recruited using the ORSEE system (Greiner, 2004), and the experiment was programmed and implemented with the help of z-Tree software (Fischbacher, 2007).

In the experiment, subjects play various versions of the trust game for 30 periods. In each period, they have an endowment of 100 points,

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<sup>6</sup>We assume that  $o$  is independent from  $Z$  since (1) subsidy is provided by a third-party and (2) both players receive it.

$E = 100$ , and the sum that they send is tripled,  $m = 3$ . The experiment is subdivided into three blocks of 10 periods each. The first and the third blocks are the same for all subjects – they face the standard trust game. However, in the second block, subjects play different versions of the trust game depending on the treatment to which they are randomly assigned: SUBLOW, SUBHIGH, SUGGEST, CONTROL.

In the second block of the SUBLOW treatment, subjects can gain a subsidy of 20 points,  $Z = 20$ , if the amount sent by the trustor exceeds a (low) threshold of 30,  $T = 30$ . See the game flow for the SUBLOW treatment in Figure 3.

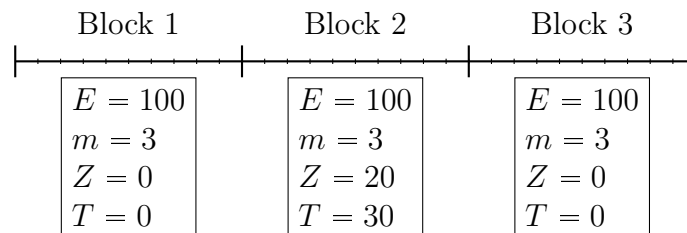


Figure 3: SUBLOW treatment parameters.

The SUBHIGH treatment differs from the SUBLOW treatment only in the threshold level: To gain the subsidy the trustor needs to send not less than 70,  $T = 70$ . See Figure 4.

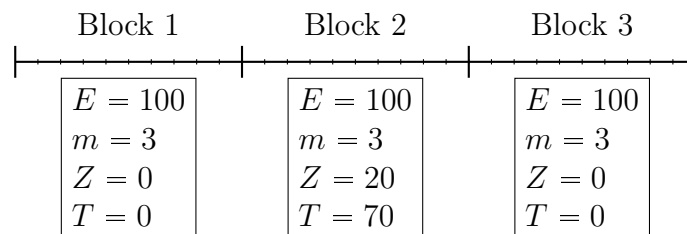


Figure 4: SUBHIGH treatment parameters.

In the SUGGEST treatment – the case of targeting policy – the subsidy is absent in all blocks, but in block 2 it is suggested by the experimenter to send not less than 70, so  $T = 70$  (like in SUBHIGH treatment). See Figure 5.

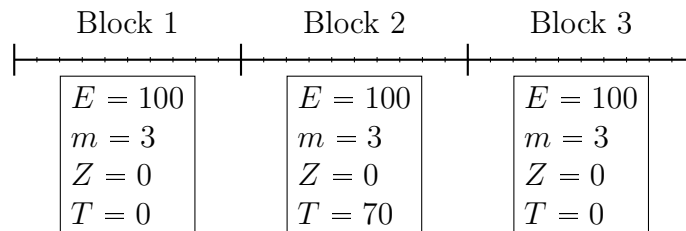


Figure 5: SUGGEST treatment parameters.

In the CONTROL treatment, the standard trust game, without any subsidies or suggestions, is played for all three blocks. See Figure 6.

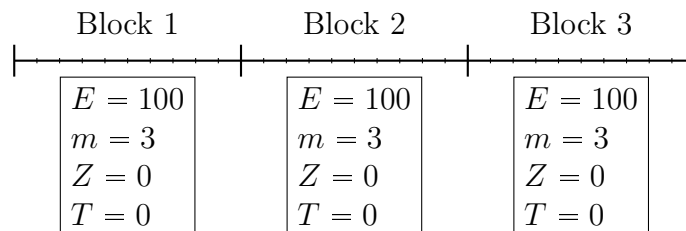


Figure 6: CONTROL treatment parameters.

We run all four treatments within the same session to control for the session specific effects. Subjects are randomly assigned to the treatment and to the role of the trustor or trustee. They keep their role throughout the whole experiment and are randomly matched with the other players from the same treatment in each period of the experiment (stranger matching design).<sup>7</sup> We keep the roles constant and use stranger matching since this design represents in our view a situation of repeated but independent decisions of the university to engage in spin-off activities.

The subjects privately receive payments at the end of the experiment according to the points they gained in one randomly selected period of the game.<sup>8</sup> Points are converted to Euros at the rate of 10 points for €0.35. Including a participation fee of €2.50, the subjects earned on average €6.81 with minimum €2.5 and maximum €15.5.

Table 1 summarizes the descriptive data about the subjects and their perception of the experiment obtained through the questionnaire given at the end of each experimental session. We almost perfectly balanced the

<sup>7</sup>Though the order of matching is random, it is identical in all four treatments within the same session. That allows us to reduce the potential effects resulting from the history of the interaction.

<sup>8</sup>We use this scheme to avoid the endowment effect. See Azrieli et al. (2012) for the analysis of incentive schemes in experiments.

sample on gender across the experiment (ratio of female participants: 0.49) and across sessions (ratio of female participants per session: 0.47, 0.5, 0.5, 0.47, 0.47, 0.53, 0.5). Also, we covered a wide range of age groups from 18 to 48 though most of the participants are relatively young (median age: 23.5).

As concerns the complexity of the experiment, subjects report a fairly high understanding of instructions with average value of 4.14 on a scale from 1 to 5 and the task difficulty as low, with mean 2.27 on a scale from 1 to 10.

Table 1: Participants characteristics

| Statistic            | N   | Mean   | St. Dev. | Min | Max |
|----------------------|-----|--------|----------|-----|-----|
| Age                  | 224 | 24.147 | 4.040    | 18  | 48  |
| Share of Females (♀) | 224 | 0.491  | 0.501    | 0   | 1   |
| Exp. Interesting     | 224 | 2.536  | 1.249    | 1   | 5   |
| Exp. Length          | 224 | 2.304  | 0.871    | 1   | 5   |
| Exp. Understandable  | 224 | 4.143  | 1.174    | 1   | 5   |
| Task difficulty      | 224 | 2.268  | 1.556    | 1   | 8   |

## 5 Results

### 5.1 Descriptive Analysis of Trust and Trustworthiness.

To assess subject's behavior, we first compare the average amount sent in each round across the treatments. Figure 7 plots the average amounts sent over the game. The average amount sent across all the treatments in block 1 is similar to what other studies find<sup>9</sup> and equals to 40.24. From visual inspection, no evident difference in trust level shows up in block 1 across the four treatments. This is to be expected since subjects play the same standard trust game in all four treatments.

Now let's consider block 2. It is clear that subjects send more on average in the treatment SUGGEST than in any other treatment. This is especially evident if one compares the treatment SUGGEST (mean: 54.84) to CONTROL (mean: 40.22). One can also observe that the curve of the

<sup>9</sup>See Johnson and Mislin (2011) for a meta-analysis of experiments based on the trust games.

average amount sent in the treatment with suggestion is always above the similar curve for the treatments with subsidy. However, the plot does not show the difference between the amount sent in treatments with subsidy and the control treatment.

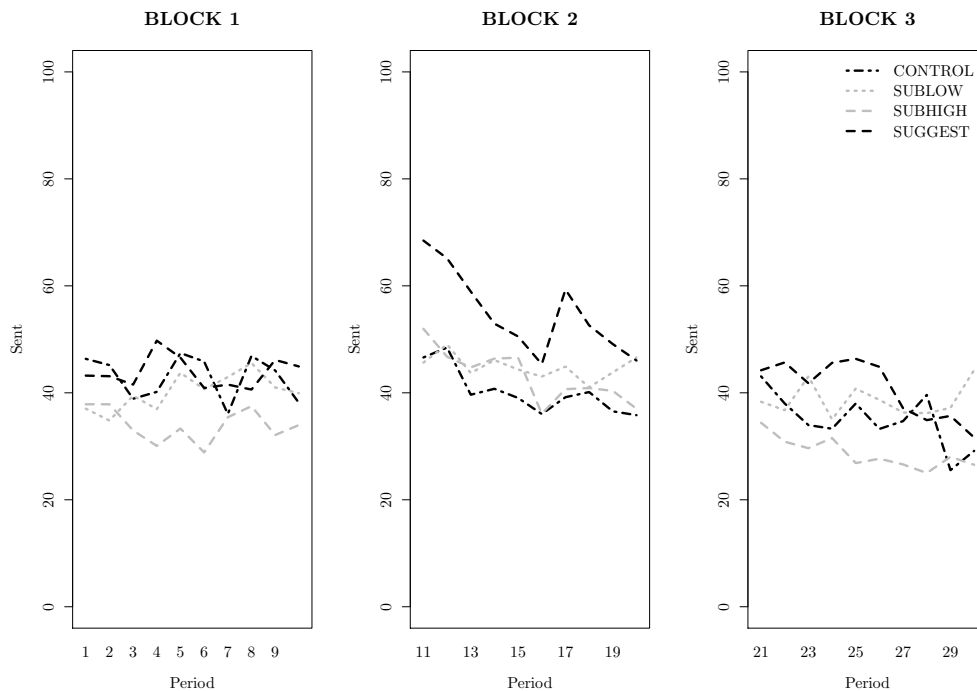


Figure 7: Average amount sent by treatment.

An interesting pattern emerges after the policy intervention. In block 3 the average amount sent in the SUGGEST treatment continues to exceed the corresponding value in the CONTROL treatment until the last periods of the game. On the contrary, the amount sent in the SUBHIGH treatment is lower than for CONTROL. The average sending in SUBLOW treatment is similar to the corresponding value in CONTROL treatment.

To have a more clear picture of the difference between the treatments, we plot the cumulative distribution functions (CDF) for each of the three blocks (see Figure 8). The CDFs indicate the proportion of cases where the amount sent is smaller than a certain value, allowing us to have a detailed view of the distribution of the amount sent.

Again, we do not see any substantial difference between treatments in block 1 but we observe a very different shape of the distributions in block 2. One can easily identify discontinuities in correspondence to the values of the low threshold ( $T = 30$ ) for the treatment SUBLOW, the high threshold

( $T = 70$ ) for the treatment SUBHIGH and the suggested amount to be sent ( $T = 70$ ) for the treatment SUGGEST. Indeed, we observe changes related to the policy intervention.

Interestingly, we see very different distributions of the amount sent for the SUBHIGH and SUGGEST treatments if we look at the values that exceed 70 (the high threshold level or the suggested amount to send). While in the SUGGEST treatment subjects do not simply send the minimal level suggested, but continue to send higher values as well, in the SUBHIGH treatment almost no one provide contributions that are higher than that required for the subsidy. This pattern can be potentially explained by a crowding out effect and we will discuss it in more details in section 5.4.

As concerns block 3, one can observe that the curve of the cumulative distribution function for SUBHIGH treatment lies below the one of the CONTROL treatment and, on the contrary, the curve for SUGGEST treatment is above the one of CONTROL treatment.

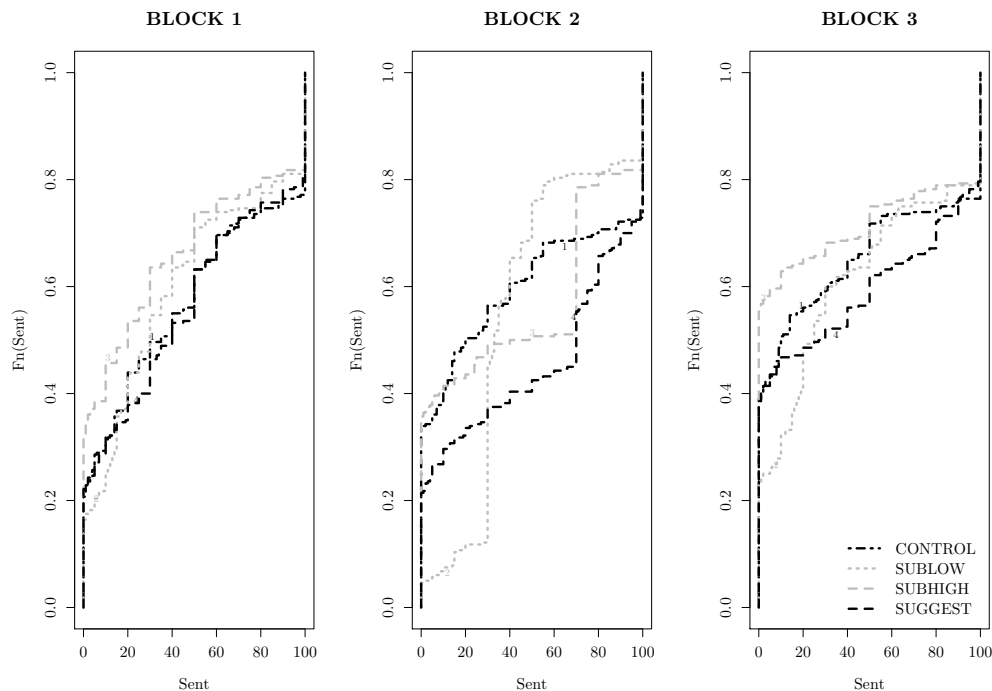


Figure 8: Cumulative distribution of amount sent by treatment.

We conclude the descriptive analysis by discussing how much player 2 (trustee) sends back. To control for the fact that the amount the subjects can send back depends on the amount received, we calculate the ratio between the amount sent back by player 2 and the amount received by the

same player - the trustworthiness rate,  $r = \frac{R}{3 \cdot s}$ . As expected, we do not observe any difference in trustworthiness between treatments (see Figure 9). The stability of trustworthiness across the treatments makes it possible to focus on the aim of our study, the analysis of the effects of external interventions on trust.

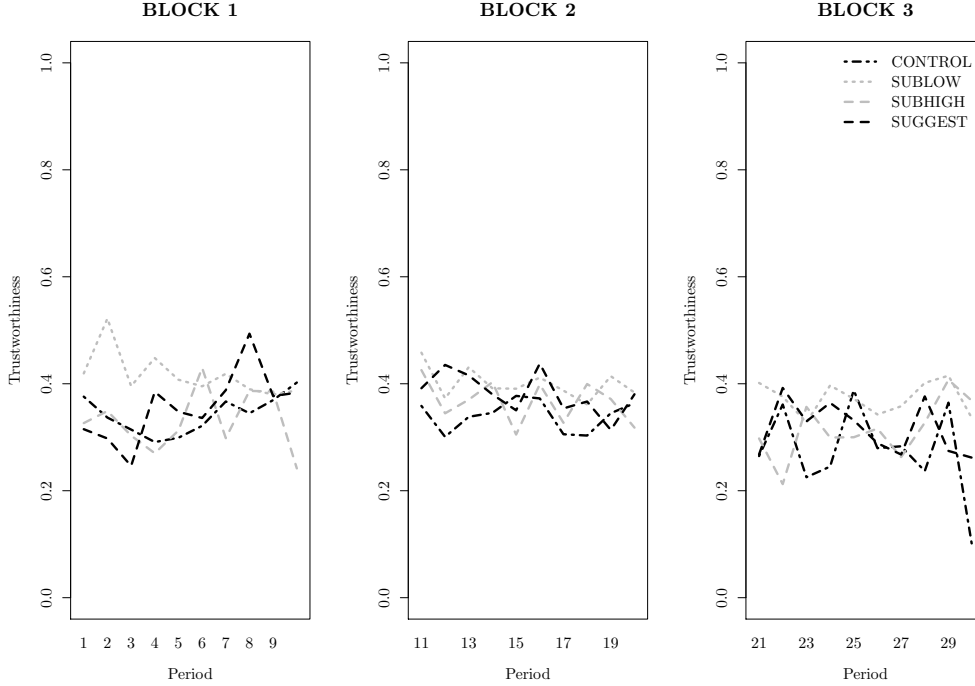


Figure 9: Average trustworthiness by treatment.

## 5.2 Regression Analysis of Trust and Trustworthiness

To assess significance of our results, we provide a regression analysis using a mixed effects model. We estimate the difference in amount sent, in the trust level across treatments by running the following regression:

$$s = \beta_0 + \beta_{SG}SUGGEST + \beta_{SL}SUBLOW + \beta_{SH}SUBHIGH + v_i + \epsilon_{i,t}, \quad (11)$$

where  $SUGGEST$ ,  $SUBLOW$ ,  $SUBHIGH$  are dummy variables that are equal to 1 for the corresponding treatments.  $v_i$  is the random effect for subject  $i$  and  $\epsilon_{i,t}$  is the error term for subject  $i$  in period  $t$ . The results are reported in Table 2.

In line with expectations and the observed pattern in Figure 7 we do not find a significant difference at any conventional level in the first ten



Table 2: Determinants of Sending by five periods – estimation of equation 11

|                     | Sent (s)         |                  |                  |                  |                  |                  |
|---------------------|------------------|------------------|------------------|------------------|------------------|------------------|
|                     | Periods          |                  |                  |                  |                  |                  |
|                     | 1-5              | 5-10             | 11-15            | 16-20            | 21-25            | 26-30            |
| SUGGEST             | 1.2<br>(8.4)     | 0.6<br>(10.0)    | 16.3*<br>(8.7)   | 12.9<br>(9.7)    | 7.5<br>(10.1)    | 4.4<br>(10.0)    |
| SUBLOW              | -5.2<br>(8.4)    | -0.2<br>(10.0)   | 2.9<br>(8.7)     | 6.3<br>(9.7)     | 1.5<br>(10.1)    | 6.1<br>(10.0)    |
| SUBHIGH             | -9.2<br>(8.4)    | -8.7<br>(10.0)   | 4.4<br>(8.7)     | 1.5<br>(9.7)     | -6.6<br>(10.1)   | -5.7<br>(10.0)   |
| Constant            | 43.6***<br>(5.9) | 42.2***<br>(7.1) | 42.9***<br>(6.2) | 37.5***<br>(6.9) | 37.3***<br>(7.2) | 32.5***<br>(7.1) |
| Observations        | 560              | 560              | 560              | 560              | 560              | 560              |
| Akaike Inf. Crit.   | 5,207.3          | 5,079.5          | 5,273.9          | 5,118.6          | 5,085.7          | 5,229.5          |
| Bayesian Inf. Crit. | 5,233.3          | 5,105.5          | 5,299.8          | 5,144.6          | 5,111.7          | 5,255.4          |

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

periods. The behavior should not differ since there is no intervention in the first ten periods (block 1).

Now let's consider the effect of the intervention. We observe that subjects send significantly more in the SUGGEST treatment than in the CONTROL treatment during the first 5 periods of block 2 ( $p = 0.064$ ;  $\beta_{SG} = 16.3$ ).<sup>10</sup> During the next 5 periods of block 2 this difference remains positive, however, it is no longer significant ( $p = 0.186$ ;  $\beta_{SG} = 12.943$ ). Thus, we conclude that the targeting policy reaches its goal and positively affects the level of sending though only in the short-run.

As concerns subsidy-policy, its effect is less evident. We cannot reject the null-hypothesis that the average amount sent in the treatments with subsidies is the same as the average amount sent in the control treatment neither in the first five periods of block 2 (SUBHIGH:  $p = 0.742$ ;  $\beta_{SH} = 2.871$ ; SUBLOW:  $p = 0.616$ ;  $\beta_{SL} = 4.386$ ) nor for the next five periods (SUBHIGH:  $p = 0.516$ ;  $\beta_{SH} = 6.336$ ; SUBLOW:  $p = 0.88$ ;  $\beta_{SL} = 1.471$ ). Put it differently, we do not find an evidence that subsidy policy is an effective mean to promote trustful behavior in the short-run.

In the last ten periods of the game we do not find any significant post-intervention effects. The amount sent in the control treatment does

<sup>10</sup>Here and after the p-values for the linear models are obtained using the approximation of Kenward and Roger (1997).

not significantly differ from the one in any other treatment.<sup>11</sup> We observe, however, that the coefficient associated with the dummy for SUGGEST treatment is positive and larger than in the first periods of the game. This result suggests that there can be a long-lasting effect of the targeting policy, though a further investigation is necessary.

We conclude this section by analyzing the evolution of trustworthiness  $r$ . Similarly to (11) we estimate the following regression:

$$r = \beta_0 + \beta_{SG}^r SUGGEST + \beta_{SL}^r SUBLOW + \beta_{SH}^r SUBHIGH + s + v_i + \epsilon_{i,t}, \quad (12)$$

In line with the theoretical predictions we do not find a significant difference over the entire experiment in trustworthiness rate between CONTROL, SUBHIGH, and SUGGEST treatments (see Table 7 in appendix B.1). Trustworthiness is significantly different during the first five periods ( $p = 0.047$ ;  $\beta_{SL}^r = 0.11$ ) and the last five periods of the game for the SUBLOW treatment ( $p = 0.071$ ;  $\beta_{SL}^r = 0.124$ ). This difference might be driven by subjects idiosyncratic characteristics. To avoid interpretation of potentially biased results in the SUBLOW treatment we focus on the CONTROL, SUBHIGH and SUGGEST treatments, though we report the analysis of subjects behavior in SUBLOW treatment as well.

It is especially interesting to see no difference in trustworthiness between the treatments with subsidy and control during the intervention period: The subjects that are exposed to subsidy still do not significantly change their behavior. It indirectly points out that unconditional subsidy does not produce crowding out effect.

### 5.3 Net Payoff

Now, we consider how the reaction on different policies is reflected in the variation of payoffs. Specifically, we evaluate the effect of each policy on the average net payoff  $\pi_N$ , that is, the difference between the subject's payoff and the value of the subsidy (s)he gets:  $\pi_N = \pi - Z$ . We subtract the value of a subsidy to account for the costs of the third party. The following mixed-effect model is estimated:

$$\pi_N = \beta_0 + \beta_{SG}^\pi SUGGEST + \beta_{SL}^\pi SUBLOW + \beta_{SH}^\pi SUBHIGH + P + v_i + \epsilon_{i,t}, \quad (13)$$

where  $SUGGEST$ ,  $SUBLOW$ ,  $SUBHIGH$  are dummy variables that are equal to 1 for the corresponding treatments.  $P$  is a dummy variable that is equal to 1 if the player is a trustor and 0 if the player is a trustee.  $v_i$

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<sup>11</sup>We as well do not find any significant difference comparing each of the treatments to each other.

is the random effect for subject  $i$  and  $\epsilon_{i,t}$  is the error term for subject  $i$  in period  $t$ .

Table 3: Determinants of Net Payoff ( $\pi_N$ ) by five periods – estimation of equation 13

|                     | <i>Dependent variable:</i> |                   |                   |                   |                   |                   |
|---------------------|----------------------------|-------------------|-------------------|-------------------|-------------------|-------------------|
|                     | Net Payoff ( $\pi_N$ )     |                   |                   |                   |                   |                   |
|                     | 1-5                        | 5-10              | 11-15             | 16-20             | 21-25             | 26-30             |
| SUGGEST             | 1.2<br>(6.7)               | 0.6<br>(6.9)      | 16.3**<br>(8.1)   | 12.9*<br>(7.7)    | 7.5<br>(8.3)      | 4.4<br>(8.2)      |
| SUBHIGH             | -5.2<br>(6.7)              | -0.2<br>(6.9)     | 2.9<br>(8.1)      | 6.3<br>(7.7)      | 1.5<br>(8.3)      | 6.1<br>(8.2)      |
| SUBLOW              | -9.2<br>(6.7)              | -8.7<br>(6.9)     | 4.4<br>(8.1)      | 1.5<br>(7.7)      | -6.6<br>(8.3)     | -5.7<br>(8.2)     |
| Player (P)          | -67.8***<br>(4.7)          | -59.3***<br>(4.9) | -76.9***<br>(5.8) | -67.9***<br>(5.4) | -69.3***<br>(5.9) | -65.9***<br>(5.8) |
| Constant            | 177.5***<br>(5.3)          | 171.9***<br>(5.4) | 181.3***<br>(6.4) | 171.5***<br>(6.1) | 171.9***<br>(6.6) | 165.4***<br>(6.5) |
| Observations        | 1,120                      | 1,120             | 1,120             | 1,120             | 1,120             | 1,120             |
| Akaike Inf. Crit.   | 12,329.4                   | 12,493.2          | 12,459.2          | 12,474.6          | 12,581.9          | 12,671.8          |
| Bayesian Inf. Crit. | 12,364.6                   | 12,528.4          | 12,494.3          | 12,509.7          | 12,617.0          | 12,707.0          |

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

As expected we find a significant increase in net payoffs during the first five periods of targeting policy ( $p = 0.046$ ;  $\beta_{SG}^\pi = 16.3$ ) as well as during the next five ( $p = 0.092$ ;  $\beta_{SG}^\pi = 12.943$ ). On the contrary, we still do not find significant effect of subsidy policy: The subsidy policy is ineffective both during the first five periods of intervention (SUBHIGH:  $p = 0.59$ ;  $\beta_{SH}^\pi = 4.386$ ; SUBLOW:  $p = 0.724$ ;  $\beta_{SL}^\pi = 2.871$ ) and during the next five (SUBHIGH:  $p = 0.848$ ;  $\beta_{SH}^\pi = 1.471$ ; SUBLOW:  $p = 0.409$ ;  $\beta_{SL}^\pi = 6.336$ ). To shed light on the reasons of these results, we provide further analysis in the next subsection.

## 5.4 Crowding Out

We wish to understand the potential cause of inefficiency of subsidy policy. To do that we focus on the distribution of the amount sent in treatments with different policy but with identical threshold level: SUBHIGH and SUGGEST.

At first we look at the subject's general propensity to follow the subsidy and the targeting policy. We compare the probability that subjects send an amount that is greater or equal to 70 in the SUBHIGH and SUGGEST treatments as opposed to CONTROL treatment. We do this by estimating the following regression:

$$Pr(s \geq 70) = \mathcal{L}(\beta_0 + \beta_{SG}^{\geq} SUGGEST + \beta_{SH}^{\geq} SUBHIGH + v_i), \quad (14)$$

where  $\mathcal{L}$  is a standard logistic function. The results are reported in Table 4.

Table 4: Determinants  $Pr(s \geq 70)$  by five periods – estimation of equation 14

|                     | $Pr(s \geq 70)$  |                  |                 |                  |                  |                  |
|---------------------|------------------|------------------|-----------------|------------------|------------------|------------------|
|                     | Periods          |                  |                 |                  |                  |                  |
|                     | 1-5              | 5-10             | 11-15           | 16-20            | 21-25            | 26-30            |
| SUGGEST             | 0.8<br>(1.6)     | -0.02<br>(1.7)   | 3.0**<br>(1.2)  | 5.4**<br>(2.3)   | 0.8<br>(1.8)     | 0.1<br>(1.6)     |
| SUBHIGH             | -0.2<br>(1.5)    | -0.6<br>(1.8)    | 2.1*<br>(1.1)   | 4.2**<br>(2.0)   | -0.4<br>(1.8)    | -0.4<br>(1.6)    |
| Constant            | -6.0***<br>(2.0) | -8.4***<br>(1.5) | -1.9**<br>(0.8) | -5.3***<br>(1.6) | -8.6***<br>(1.5) | -7.7***<br>(1.5) |
| Observations        | 420              | 420              | 420             | 420              | 420              | 420              |
| Akaike Inf. Crit.   | 338.9            | 266.9            | 399.4           | 329.1            | 274.3            | 285.8            |
| Bayesian Inf. Crit. | 355.1            | 283.1            | 415.6           | 345.3            | 290.5            | 301.9            |

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Of course, we find a significant increase in propensity to follow the targeting policy for the first five ( $p = 0.0118$ ;  $\beta_{SG}^{\geq} = 3.02$ ;  $e^{\beta_{SG}^{\geq}} = 20.56$ ) as well as for the next five periods of block 2 ( $p = 0.0167$ ;  $\beta_{SG}^{\geq} = 5.4$ ;  $e^{\beta_{SG}^{\geq}} = 222.45$ ). It is, however, more surprising to observe that subjects are significantly more likely to send the required amount during the subsidy policy as well (for periods 10–15:  $p = 0.0632$ ;  $\beta_{SH}^{\geq} = 2.09$ ;  $e^{\beta_{SH}^{\geq}} = 8.04$ ; for periods 16–20:  $p = 0.0402$ ;  $\beta_{SH}^{\geq} = 4.2$ ;  $e^{\beta_{SH}^{\geq}} = 67$ ).

This result is puzzling since we do not observe that subjects send significantly more on average in the SUBHIGH than in the CONTROL treatment in block 2 (see Table 2 in section 5.2).<sup>12</sup> We can partially explain it by the fact that subjects' propensity to follow the policy tends to be lower in case of the SUBHIGH than in the SUGGEST treatment (for periods 10–15:  $\beta_{SH}^{\geq} = 2.09 < \beta_{SG}^{\geq} = 3.02$ ; for periods 16–20:  $\beta_{SH}^{\geq} = 4.2 < \beta_{SG}^{\geq} = 5.4$ ). Thus, given the sample size, we may not capture the effect directly.

The observed pattern points out that subsidy policy significantly affects the subjects' behavior but it is not that effective as the targeting policy because subjects avoid to follow the subsidy policy. This explanation can be partially accepted, however, one needs to compare whether the propensity to follow the policy is, indeed, significantly lower in case of subsidy than in case of suggestion. To do that we estimate the following regression

<sup>12</sup>As well as given that we do not observe significant growth in net payoffs during the subsidy policy (see Table 3 in section 5.3).

using the SUBHIGH treatment as a reference category:

$$Pr(s \geq 70) = \mathcal{L}(\beta_0 + \beta_{SG}^{\geq} SUGGEST + v_i) \quad (15)$$

Nonetheless we do not find a significant difference in propensity to follow the policy between the SUGGEST and SUBHIGH treatments neither in the first five periods ( $p = 0.3963$ ;  $\beta_{SG}^{\geq} = 0.89$ ;  $e^{\beta_{SG}^{\geq}} = 2.43$ ) nor in the next five periods of block 2 ( $p = 0.6082$ ;  $\beta_{SG}^{\geq} = 0.73$ ;  $e^{\beta_{SG}^{\geq}} = 2.08$ ). The results are reported in Table 8 in Appendix B.2. It suggests that another source of inefficiency is possibly at work and to find it we have a closer look at the distributions of the sendings in the SUBHIGH and SUGGEST treatments.

We have mentioned in Section 5.1 that the distribution of the sendings is different for the SUBHIGH and SUGGEST treatments in block 2. Subjects tend to send not more than the minimal amount 70 required to get the subsidy in the SUBHIGH treatment, while in the SUGGEST treatment the subjects also send more than the minimal level suggested (see Figure 8). If this difference is significant it explains why the effect of the subsidy policy is not as large as the effect of the targeting policy.

To assess the significance of the observed disparity we evaluate whether the probabilities to send an amount that is greater than 70 or equal to 70 are different between the SUBHIGH and SUGGEST treatments. We estimate the following two logistic regressions using the SUBHIGH treatment as a reference category:

$$Pr(s = 70) = \mathcal{L}(\beta_0 + \beta_{SG}^= SUGGEST + v_i) \quad (16)$$

$$Pr(s > 70) = \mathcal{L}(\beta_0 + \beta_{SG}^{>} SUGGEST + v_i) \quad (17)$$

We report the results in the Tables 5 and 6. One can see that the probability of sending exactly 70 is significantly lower in the SUGGEST treatment as compared to the SUBHIGH treatment during the first five periods of block 2 ( $p = 0.0234$ ;  $\beta_{SG}^= = -1.71$ ;  $e^{\beta_{SG}^=} = 0.18$ ). On the contrary, the probability of sending more than 70 is significantly higher in the SUGGEST treatment (than in the SUBHIGH treatment) also during the first five periods of block 2 ( $p = 0.0254$ ;  $\beta_{SG}^{>} = 3.44$ ;  $e^{\beta_{SG}^{>}} = 31.29$ ).

Moreover, applying the non-parametric exact paired Wilcoxon test across aggregated averages over the sessions, we reject the null-hypothesis that there is no difference between the SUGGEST and SUBHIGH treatments in probability to send exactly 70 ( $p = 0.0076$ ) and more than 70 ( $p = 0.046$ ) during the first five periods of block 2.<sup>13</sup>

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<sup>13</sup>We estimate the exact paired Wilcoxon test based on the Shift Algorithm by Streitberg and Röhmel (1986).

That is, in the SUGGEST treatment subjects tend to send more than 70 and, hence, contribute to the growth of the average amount sent. However, in the SUBHIGH treatment subjects tend to fulfill the requirement to get the subsidy but not to send more, diminishing the average level of contribution. Thus, the specific reaction on the subsidy policy decreases its effectiveness.

Table 5: Determinants of  $Pr(s = 70)$  by five periods – estimation of equation 16

|                     | $Pr(s = 70)$     |                  |                  |                 |                  |                   |
|---------------------|------------------|------------------|------------------|-----------------|------------------|-------------------|
|                     | Periods          |                  |                  |                 |                  |                   |
|                     | 1-5              | 5-10             | 11-15            | 16-20           | 21-25            | 26-30             |
| SUGGEST             | 1.1<br>(1.2)     | 0.7<br>(1.2)     | -1.7**<br>(0.8)  | -1.7<br>(1.6)   | -0.7<br>(1.2)    | -0.51<br>(2.2)    |
| Constant            | -4.9***<br>(1.0) | -4.9***<br>(1.0) | -1.6***<br>(0.5) | -5.3**<br>(2.6) | -4.2***<br>(0.7) | -11.6***<br>(3.5) |
| Observations        | 280              | 280              | 280              | 280             | 280              | 280               |
| Akaike Inf. Crit.   | 46.9             | 38.8             | 234.8            | 186.3           | 38.8             | 27.6              |
| Bayesian Inf. Crit. | 57.8             | 49.7             | 245.7            | 197.2           | 49.7             | 38.5              |

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

Table 6: Determinants of  $Pr(s > 70)$ , in block 2 – estimation of equation 17

|                     | <i>Dependent variable:</i> |                  |                  |                  |                  |                  |
|---------------------|----------------------------|------------------|------------------|------------------|------------------|------------------|
|                     | $Pr(s > 70)$               |                  |                  |                  |                  |                  |
|                     | 1-5                        | 5-10             | 11-15            | 16-20            | 21-25            | 26-30            |
| SUGGEST             | 0.7<br>(1.1)               | 0.4<br>(1.9)     | 3.4**<br>(1.5)   | 2.0<br>(2.0)     | 1.2<br>(1.8)     | 0.7<br>(1.7)     |
| Constant            | -3.2***<br>(1.1)           | -9.3***<br>(1.7) | -3.7***<br>(1.3) | -8.5***<br>(1.8) | -8.8***<br>(1.7) | -8.8***<br>(1.6) |
| Observations        | 280                        | 280              | 280              | 280              | 280              | 280              |
| Akaike Inf. Crit.   | 243.1                      | 158.7            | 241.1            | 193.9            | 187.7            | 177.5            |
| Bayesian Inf. Crit. | 254.0                      | 169.6            | 252.0            | 204.8            | 198.6            | 188.4            |

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01

## 6 Discussion and Conclusion

In this section, we assess our contribution to the literature. We highlight the main results and provide a short discussion of the efficiency of the studied external interventions and the mechanisms acting behind them.

Our analysis falls under the broad rubric of studies on incentives and social preferences. It combines an exercise in decomposition of preferences to uncover sources of crowding-out with an attempt to account for the long-run detrimental effects of incentives.

We develop a model that predicts that the policy that involves monetary incentives can be ineffective since this type of incentives crowd-out other regarding preferences if subjects comply with the policy. We assume that preferences are endogenous (Bowles, 1998) – the preference once learned stays unchanged for some time. Therefore, the monetary-based policy that eradicates social preferences negatively affects the subjects' pro-social behavior after the intervention. On the contrary, the policy that uses non-monetary incentives is effective in the short-run and does not have detrimental consequences in the long-run because it does not influence other regarding preferences.

The experimental results, indeed, show that the non-monetary incentives are an effective tool to foster pro-social behavior, namely, trustful behavior in the short-run, while there is no evidence of detrimental effects of this type of incentives in the long-run. In turn, monetary incentives do not show their effectiveness in the short- as well as in the long-run though the policy significantly affects the subjects' behavior during the intervention. To interpret this fact we turn to the taxonomy of incentive effects on preferences provided by Bowles and Polania-Reyes (2012).

According to their taxonomy, there are three mechanisms linking interventions and preferences: “bad news” – incentives provide information about interests of a principal; “control aversion” – incentives jeopardize self-determination; “moral disengagement” – incentives activate a switch from pro-social to own payoff maximization mode of thought. We do not consider here the first one (“bad news”) since the incentives are provided by the third-party and, hence, should not affect the subjects' behavior. However, the last two – “control aversion” and “moral disengagement” – can explain the specific pattern of subjects reaction on the subsidy policy.

Subjects react to the monetary policy but (1) their propensity to follow this policy is low and (2) those who follow the policy send the minimal amount required to get the subsidy. We attribute the low propensity to follow the policy to the mechanism of “control aversion”: Subjects perceive the policy as controlling and avoid following it. The “moral disengagement” can explain the fact that subjects send mostly the minimal amount: They switch their way of thinking to own-payoff maximization, thence, if they decide to follow the policy they simply minimize their costs by sending the minimal amount.

As concerns the post-intervention effect of the policies, despite the fact

that we do not find a significant difference between treatments after policy interventions, we observe that subjects tend to send a high amount after the targeting policy. This is an interesting observation since it suggests that a targeting policy can have a potential long-lasting effect. Nevertheless, further research is needed to test this observation.

It is also interesting to observe that the trustworthiness rate is not affected during the intervention as well as it does not change afterwards. On the one hand, this goes in line with theoretical expectations – the trustee’s behavior should remain the same since the policies incentivize only trustors. On the other hand, given that trustees also receive subsidies, this fact suggests that the presence of a subsidy is insufficient to crowd out other-regarding preferences. It is rather likely that the crowding out occurs when the monetary incentives are conditioned on a certain behavior.

To sum up and conclude, in this study, we aim to understand how subsidy and targeting policies affect an investment decision. We employ a multi-period trust (investment) game where we introduce an external intervention either in form of subsidy or suggestion and analyze the level of trustful behavior during and after the intervention.

We find that targeting is an effective instrument to promote trustful behavior whereas subsidy policy is not effective both in the short- and long-run: Subjects follow the targeting policy and send even more than minimal level requested, while under the subsidy policy they exhibit low propensity to follow the policy and send mostly the minimal amount needed to get the subsidy. We therefore recommend the targeting policy as one of the instruments to foster trustful behavior.

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## **A Instructions**

### **A.1 Player 1, Trustor.**

#### **Welcome to the experiment!**

Thank you very much for participating. We hope that you feel comfortable. We ask you to remain quiet and do not communicate with any other player. Please understand that in case you communicate with other players we will have to exclude you from the experiment without payment. If you have any questions please raise your hand and wait for the experimenter to come to you.

We guarantee that all information collected during the experiment undergoes a strict anonymity process. It ensures anonymity among players and that you stay anonymous to the experimenter.

During the experiment you will see information about other players. We have ensured that you cannot identify them personally as well as they cannot identify you.

The experiment is on decision-making. Your earnings will depend partly on your decisions and partly on the decisions of other players. You will have to make one decision in each round of a simple game which consists of 30 rounds.

In each round of the game the earnings will be calculated in points. At the end of the experiment **one round** will be randomly chosen. The points gained during this round will be converted to Euros with the following rate:

$$10 \text{ points} = 0.35 \text{ Euro}$$

In addition, you will receive 2.50 euro as a compensation for showing up on time. The game you will play is divided **into three blocks** (A, B and C), with 10 rounds in each block.

In each round of any block you will be matched with another randomly chosen player among other participants. There will be a new random pair each round.

The information about your previous decisions will not be revealed to other players at any round of the experiment.

In each round you and the other player both will be endowed with 100 points. **You can send any amount** to the other player. **Each point you send is tripled.** The other player will decide how many points to send back to you and how many points to keep (from zero to the tripled sum you sent).

[For the SUBHIGH and SUBLOW treatment we add the following paragraph]

Also, in some blocks if you send **not less than a certain minimum**, you and the other player will receive **an additional payment**. The amount of the **additional payment** and the required minimum sent to receive it will be specified in the beginning of each block.

[For the SUGGEST treatment we add the following paragraph]

In some blocks it will be suggested to send not less than a certain amount. The amount suggested is specified at the beginning of each block.

## A.2 Player 2, Trustee.

### Welcome to the experiment!

Thank you very much for participating. We hope that you feel comfortable. We ask you to remain quiet and do not communicate with any other player. Please understand that in case you communicate with other players we will have to exclude you from the experiment without payment. If you have any questions please raise your hand and wait for the experimenter to come to you.

We guarantee that all information collected during the experiment

undergoes a strict anonymity process. It ensures anonymity among players and that you stay anonymous to the experimenter.

During the experiment you will see information about other players. We have ensured that you cannot identify them personally as well as they cannot identify you.

The experiment is on decision-making. Your earnings will depend partly on your decisions and partly on the decisions of other players. You will have to make one decision in each round of a simple game which consists of 30 rounds.

In each round of the game the earnings will be calculated in points. At the end of the experiment **one round** will be randomly chosen. The points gained during this round will be converted to Euros with the following rate:

$$10 \text{ points} = 0.35 \text{ Euro}$$

In addition, you will receive 2.50 euro as a compensation for showing up on time. The game you will play is divided **into three blocks** (A, B and C), with 10 rounds in each block.

In each round of any block you will be matched with another randomly chosen player among other participants. There will be a new random pair each round.

The information about your previous decisions will not be revealed to other players at any round of the experiment.

In each round you and the other player both will be endowed with 100 points. You will receive some amount of points from the other player. **Each point sent by the other player is tripled. You can decide** how many points to send back to him and how many points to keep (from zero to the tripled sum of points the other player sent).

[For the SUBHIGH and SUBLOW treatment we add the following paragraph.]

Also, in some blocks if the other player sends **not less than a certain minimum**, you and the other player will receive **an additional payment**. The amount of **the additional payment** and the required minimum sent to receive it will be specified in the beginning of each block.

[For the SUGGEST treatment we add the following paragraph]

In some blocks, it will be suggested to other player to send not less than a certain amount. The amount suggested is specified at the beginning of each block.

## B Additional Estimations

### B.1 Trustworthiness

Table 7: Determinants of trustworthiness by five periods – estimation of equation 12

|                     | Trustworthiness ( $r$ ) |                     |                     |                     |                     |                    |
|---------------------|-------------------------|---------------------|---------------------|---------------------|---------------------|--------------------|
|                     | Periods                 |                     |                     |                     |                     |                    |
|                     | 1-5                     | 5-10                | 11-15               | 16-20               | 21-25               | 26-30              |
| SUGGEST             | −0.004<br>(0.1)         | 0.1<br>(0.1)        | 0.04<br>(0.1)       | 0.04<br>(0.1)       | 0.1<br>(0.1)        | 0.1<br>(0.1)       |
| SUBLOW              | 0.1**<br>(0.1)          | 0.05<br>(0.1)       | 0.1<br>(0.1)        | 0.1<br>(0.1)        | 0.1<br>(0.1)        | 0.1*<br>(0.1)      |
| SUBHIGH             | −0.01<br>(0.1)          | 0.01<br>(0.1)       | 0.03<br>(0.1)       | 0.04<br>(0.1)       | 0.02<br>(0.1)       | 0.1<br>(0.1)       |
| Sent ( $s$ )        | 0.002***<br>(0.000)     | 0.002***<br>(0.000) | 0.001***<br>(0.000) | 0.001***<br>(0.000) | 0.001***<br>(0.000) | 0.001**<br>(0.000) |
| Constant            | 0.3***<br>(0.04)        | 0.3***<br>(0.05)    | 0.3***<br>(0.1)     | 0.2***<br>(0.1)     | 0.2***<br>(0.05)    | 0.2***<br>(0.1)    |
| Observations        | 459                     | 405                 | 455                 | 398                 | 371                 | 312                |
| Log Likelihood      | 57.4                    | 51.6                | 102.8               | 94.6                | 64.1                | 31.6               |
| Akaike Inf. Crit.   | −100.7                  | −89.1               | −191.7              | −175.2              | −114.3              | −49.1              |
| Bayesian Inf. Crit. | −71.8                   | −61.1               | −162.8              | −147.3              | −86.9               | −22.9              |

*Note:* \*p<0.1; \*\*p<0.05; \*\*\*p<0.01

## B.2 Probability to Follow the Policy (SUGGEST VS. SUBHIGH)

Table 8: Determinants  $Pr(s \geq 70)$  by five periods – estimation of equation 15

|                     | $Pr(s \geq 70)$  |                  |              |               |                  |                  |
|---------------------|------------------|------------------|--------------|---------------|------------------|------------------|
|                     | Periods          |                  |              |               |                  |                  |
|                     | 1-5              | 5-10             | 11-15        | 16-20         | 21-25            | 26-30            |
| SUGGEST             | 0.8<br>(1.1)     | 0.6<br>(1.8)     | 0.9<br>(1.0) | 0.7<br>(1.4)  | 1.2<br>(1.8)     | 0.5<br>(1.7)     |
| Constant            | -3.3***<br>(1.1) | -9.0***<br>(1.7) | 0.2<br>(0.7) | -0.7<br>(1.0) | -9.1***<br>(1.7) | -8.6***<br>(1.6) |
| Observations        | 280              | 280              | 280          | 280           | 280              | 280              |
| Akaike Inf. Crit.   | 248.9            | 172.1            | 280.0        | 250.6         | 185.4            | 180.9            |
| Bayesian Inf. Crit. | 259.8            | 183.0            | 290.9        | 261.5         | 196.3            | 191.8            |

Note:

\*p<0.1; \*\*p<0.05; \*\*\*p<0.01