

Can Nudges Be Transparent and Yet Effective?

Abstract

Nudges receive growing attention as an effective strategy to alter people's decisions without significantly changing economic incentives or limiting options. However, being often very subtle and covert, nudges are also criticized as unethical. By not being transparent about the intention to influence individual choice they might be perceived as limiting freedom of autonomous actions and decisions. So far, empirical research on this issue is scarce. In this study, we investigate whether nudges can be made transparent without limiting their effectiveness. For this purpose we conduct a laboratory experiment where we 'nudge' contributions to carbon emission reduction by introducing a default value. We test how different types of transparency (i.e. knowledge of the potential influence of the default, its purpose, or both) influence the effect of the default. Our findings demonstrate that the default increases contributions, and information on the potential influence combined with the purpose of the default, or just its purpose, do not significantly affect contributions. Findings are somewhat inconclusive with respect to information on the potential behavioral influence. Furthermore, we do not find evidence that psychological reactance interrelates with the influence of transparency. Generally, our findings support the policy-relevant claim that nudges (in the form of defaults) can be transparent and yet effective.

Keywords: climate protection, experiment, default, nudge, transparency,

public good

JEL: D03, H41, Q58, K23

PsycINFO classification code: 3000, 3900, 4000, 4070

1. Introduction

Nudges, a term coined by Thaler & Sunstein (2008), describe diverse instruments that utilize behavioral insights in order to affect individual behavior, without limiting options or significantly changing economic incentives. They have become an alternative to classical interventions from traditional economics, differing from the latter insofar as they affect behavior by changing context instead of cognition (Dolan et al., 2012). The recent success of this approach can be viewed as a direct consequence of conceiving individual behavior as bounded, instead of perfectly rational and selfish (Bolton & Ockenfels, 2012). Nudges are evolving into a popular form of soft regulation in various fields such as health, finance, environmental protection, etc. (Sunstein, 2014a; Alemanno & Sibony, 2015; World Bank, 2015; Lourenco et al., 2016). Despite growing popularity, use of behavioral insights in policy-making is subject to criticism (e.g. Hausman & Welch, 2010; Rebonato, 2014). One remarkable aspect of nudges is that they often influence individual behavior without being noticed by the affected subject (Dhingra et al., 2012; Hansen & Jespersen, 2013; Sunstein, 2016). This raises the concern that nudges covertly violate individual autonomy and are therefore unethical (Bovens, 2009; House of Lords Report, 2011). Thus, this form of regulation lacks the transparency that accompanies other regulatory instruments. For instance, when the government imposes a tax to reduce con-

sumption of a product (e.g. cigarettes), people are aware of this tax and can compel the government to justify it (Sunstein, 2014b). On the other hand, when the government sets an opt-out system instead of an opt-in system to promote certain behavior (e.g. organ donation) it exploits different psychological biases, often without people’s awareness (Hansen & Jespersen, 2013). Felsen et al. (2013) demonstrate in a vignette study that a significant proportion of individuals have reservations towards nudges they perceive as covert. Another recent research stream provides evidence of the intrinsic value of decision rights and autonomy (Fehr et al., 2013; Bartling et al., 2014; Owens et al., 2014).

To address this problem we investigate whether nudges can be made transparent without reducing their effectiveness. In this context, we take into account that the covert nature of nudges is often said to be essential for their effectiveness (Bovens, 2009; House of Lords Report, 2011). Also, we acknowledge that telling people the nudge is used to influence their decision potentially evokes a perceived threat to their freedom leading them to experience psychological reactance. The latter can be defined as ”the motivational state that is hypothesized to occur when a freedom is eliminated or threatened with elimination” (Brehm & Brehm, 2013, p. 37). This could not only inhibit the effect of the nudge but could even lead to the opposite effect than the one intended. Therefore, this psychological phenomenon is important when investigating the influence of transparency on the effectiveness of nudges.

To test the interrelation between nudges and transparency we conduct a laboratory experiment where subjects are asked how much they would be

willing to contribute to a climate protection fund. The nudge is a default value that aims to increase contributions. There are different mechanisms through which a default influences behavior, e.g. as a reference value and anchor (for construction of preferences), through provision of social norms or information, or through inertia (by imposing pecuniary or cognitive costs on deviating from the default). Sunstein & Reisch (2016) provide a review. Note that Cappelletti et al. (2014) provide evidence from a public good game that defaults do not work as recommendations, i.e. as information provision. We expect the default value to increase contributions through two possible ways. First, it can increase the fraction of people picking the default value. Second, it can induce people to increase their contribution towards this value. We discuss our findings in relation to the possible mechanisms in the last Section. The type of transparency that accompanies the default varies across treatments and includes either informing decision makers about its potential behavioral influence and/or informing them about its purpose to increase contributions to climate protection. We assess two different measures of psychological reactance after the experiment. Thus, we can investigate whether the influence of transparency is limited to a sub-group of participants distinct in their proneness to show psychological reactance (trait reactance). Additionally, we can test whether transparency influences the perception of a nudge as a threat for freedom of choice, and whether it functions as a source of anger (state reactance).

Recent findings from Arad & Rubinstein (2015) illustrate why our investigation of transparency and psychological reactance in the context of nudges is important. Their findings suggest that some subjects may consciously act

contrary to the encouraged action, presumably in order to protest against the intervention of the government. The authors argue that full transparency of nudges, thus, may even lead to the opposite outcome than the one intended (as opposed to simply eliminating the effectiveness of a nudge). Some people behave in a completely different way simply out of protest against being manipulated. Contrary to this argument, findings by Sunstein (2016) from a nationally representative survey in the USA show that there is widespread support for nudges, and that transparency concerning the nudge will not diminish its effectiveness.¹

To the best of our knowledge, there are three empirical studies directly relevant to our research question. Loewenstein et al. (2015), in a laboratory experiment, find no evidence that, informing subjects that they were presented with a pro-self² default option influences their effectiveness. Similarly, Kroese et al. (2016), in a field experiment, find no evidence that making subjects aware of the purpose behind a pro-self default has any effect. Steffel et al. (2016), in several hypothetical and marginally incentivized consumer-related experiments, find no evidence that stressing the potential behavioral influence of a pro-self, as well as a pro-social default impacts their effectiveness, although it affects perception by the consumer.

While existing evidence unanimously suggests the impact of transparency on effectiveness of nudges is absent, our research augments this in various

¹Reisch & Sunstein (2016) show that there is also a general support of nudges in six European countries.

²Hagman et al. (2015) divide nudges into pro-self and pro-social. While the former nudge people towards making better decisions for themselves, the latter nudge people towards behavior benefitting society as a whole.

ways. First, subjects in our experiment face a tradeoff between real monetary payoffs and real contributions to a (global) public good. By contrast, two of the previous studies employed relatively abstract and stylized environments, and did not demand subjects to make (substantial) financial tradeoffs. Although Kroese et al. (2016) investigate behavior in the field, they do neither study pro-social nudges, nor do they incorporate both types of transparency. Second, we investigate the distinct, as well as combined effect of two types of transparency on the default effect. Previous research focused exclusively on either of these two categories. However, there are reasons to expect that informing decision makers about the potential behavioral influence of a nudge has different consequences than informing them about its purpose. Third, we enrich our analysis with the concept of psychological reactance, allowing for a deeper understanding of potential channels through which transparency influences default effects. Recent research on nudges, although focusing conceptually on the role of reactance (Arad & Rubinstein, 2015; Hedlin & Sunstein, 2016), did not investigate its interaction with transparency.

Consequently, we contribute to the knowledge on the topic of transparency of nudges in various ways. First, we enable a more nuanced view on this topic by investigating two types of transparency, thus contributing to a better understanding on how transparency works and whether policy-makers can make nudges more transparent without diminishing effectiveness. Second, our experimental setup, albeit controlled, sets up a realistic context, enabling us to make more valid inferences about the impact of transparency on nudges in "the real world". Third, we widen the discussion on transparency

by investigating its connection to the concept of psychological reactance.

To preview our results, defaulted contributions are significantly higher than in the control group, even when accompanied by information regarding the purpose of the default, or the purpose and its potential influence. It is not clear, however, how sole provision of the potential influence of a default changes its effectiveness. In addition, contributions in the treatment groups (with or without transparency) do not significantly differ from each other. Finally, neither do we find evidence that trait reactance interacts with transparency, nor does data suggest that transparency changes the perception of nudges as freedom threatening or sources of anger. Therefore, our findings advocate that nudges (in the form of defaults) can be transparent and effective.

The remainder of the article is structured as follows. In Section 2 we discuss psychological reactance as a conceptual background to covert nudges, followed by derivation of behavioral predictions. We lay out the experimental design in Section 3. In Section 4 we present and analyze the results. Section 5 concludes.

2. Conceptual framework and behavioral predictions

Since Brehm (1966) introduced the theory of psychological reactance, many studies have explored this phenomenon. Social influence attempts (such as nudges) that are detected by an individual may be perceived as a threat to freedom of choice (Brehm, 1966). The elicited state of psychological reactance may result in behavioral and cognitive efforts to reestablish freedom as well as uncomfortable, hostile, aggressive, and angry feelings (Dillard

& Shen, 2005). Consequently, people may try to restore their freedom by exhibiting exactly the restricted behavior, thus, in our case, strongly deviating from the default value. In addition, they may devalue the source of threat (the initiator of the nudge), increase their liking for the restricted freedom, or counter-argue against the imposed option (Brehm, 1966; Dillard & Shen, 2005). People react in such a manner not only to obvious and direct, but also to subtle and subliminal threats (Chartrand et al., 2007).

In order to investigate whether transparency influences the effectiveness of pro-social nudges, specifically defaults, we chose the context of climate protection. With climate change being one of the major challenges faced by society on a global scale today, information-based instruments and nudges are becoming increasingly important to increase individual contributions to climate and environmental protection (Allcott & Mullainathan, 2010; Araña & León, 2013; World Bank, 2015).

One way to contribute to climate protection is to offset (parts of) one's own yearly CO₂ emissions by donating to specific charitable organizations (in the experiment, referred to as 'climate protection fund'). These organizations use donations to purchase and destroy carbon emission licenses from the European Union Emissions Trading Scheme (EU ETS).³ Buying carbon licenses is an effective way for individuals to contribute to climate protection, when compared to, e.g. electricity-saving (Perino, 2015). Therefore, individual payment for carbon license retirement is a relevant context in which the

³The EU ETS is a European market that ultimately prices carbon emissions and allows regulated industries to trade their emission rights. Buying licenses off the market increases the scarcity of emission rights, resulting in higher prices and thus increasing the incentives for regulated firms to invest in emission-reducing technology.

influence of transparency on the effectiveness of a pro-social nudge can be investigated.

Based on psychological reactance theory we expect that mentioning the potential influence of a default will evoke the most reactance and thus reduce its effectiveness. In contrast, the sole provision of the purpose, i.e. climate protection, should evoke little reactance since this induces perspective taking. In addition, it renders the positive goal of the contribution more salient. According to salience theory formulated by Bordalo et al. (2012), more salient attributes will be over-weighted in the decision process. Based on this argument, providing the purpose will work as an additional nudge and thus increase the default effect. Finally, accompanying the default with both types of information will be the most transparent form of the nudge. Due to combining the hypothesized "downside" effect of reactance and "upside" effect of the salience of the purpose of the nudge we expect the contribution level to be in between the other treatments. In sum, hypotheses concerning people's contribution decisions in the presence of the default are as follows:

H1: If participants are confronted with a default, contributions will be higher compared to when there is no default.

H2: If participants are informed that the default may have an influence on their decision, contributions will be lower compared to when they are not informed.

H3: If participants are informed of the purpose of the default, contribu-

tions will be higher compared to when they are not informed.

H4: If participants are informed of the potential influence of a default and of its purpose, contributions will be higher than with information solely on influence and lower than with information solely on purpose.

When analyzing findings with respect to psychological reactance, we hypothesize that the evaluation of a default as freedom-threatening, autonomy-decreasing, manipulative, and pressuring (perceived threat to freedom), as well as its potential to elicit negative emotions (anger) differs with respect to the types of transparency accompanying the default value. Specifically, we expect that:

H5: If participants are informed that the default may have an influence on their decision, experience of state reactance will be higher compared to when they are not informed.

We further hypothesize that trait reactance interacts with the type of transparency accompanying the default value. Specifically, we expect that:

H6: If participants are informed that the default may have an influence on their decision, the default effect for participants with higher trait reactance will be lower than for participants with lower trait reactance.

We deduce hypotheses H5 and H6 exclusively with respect to a default accompanied by information on its potential influence, because we expect this type of transparency to increase the salience of the potentially manipulative and autonomy-threatening default-characteristic. For the purpose of the default, the conceptual link to reactance is less clear. We therefore abstain from formulating specific hypotheses.

3. Experimental design

To test the hypotheses, we conducted a laboratory experiment that consisted of five experimental groups, of which one was the control group.⁴ A total of 214 students from the Erasmus University Rotterdam participated in the experiment in June 2016 using the z-tree software (Fischbacher, 2007), recruited through ORSEE. Of these, 43% were female, the average age was 22, and the majority (84%) studied economics. The experiment took place in the Econ-lab of the Erasmus School of Economics where participants were randomly assigned to separate computer terminals and were instructed not to communicate. The participants were given instruction sheets that were read aloud (see Appendix A). All participants received an endowment of 10 Euro and were asked to indicate how much (if any) of their endowment they would like to contribute to the 'climate protection fund'. The remaining amount served as their private payoff. After the experiment, they were

⁴Prior to the experiment, pilot sessions were conducted in Germany, Sweden, France and the Netherlands. The pilot session in Germany focused on developing the design, which was further improved on and tested among Master students in the Netherlands, Sweden, and Bachelor students in France. The experimental design was not identical in all these pilots. Therefore, findings from the pilot sessions are not included in data analysis.

paid according to their decisions, and contributions were used to retire real carbon licenses from the EU ETS, through 'TheCompensators*'.⁵ Subjects knew this prior to their decision.

In the control group, participants were presented with a text box where they could enter their contribution in any integer amount between 0 and 10 Euro. Neither a preselected default value for the contribution, nor any additional information were presented. In the other experimental groups, subjects encountered an 8 Euro default contribution in form of a button (see Figures A.2 - A.3 in Appendix A). They could either press this button or choose another one that stated 'Different amount'. In the latter case they were referred to another screen that contained exactly the same information but with the addition of a text box where they could insert any amount between 0 and 10 Euro. Once subjects made their decision, they received information regarding their contribution, their private payoff and the amount of CO₂ that would be retired with the contributed amount⁶. The treatment groups varied only with respect to the displayed information in the decision screen (Table 1).

After making their decision, participants answered a questionnaire mea-

⁵'TheCompensators*' is a non-profit association founded in 2006 by researchers from the Potsdam Institute for Climate Impact Research. They offer a way for individuals and firms to compensate for their emissions. With donations, they buy and retire emission rights from the EU ETS. At the end of the experiment, all participants received an email with a confirmation and a certificate of aggregate experimental donations to 'TheCompensators*'.
⁶At that time, "TheCompensators*" offered to retire licenses at a price of 5.53 Euro. Note that this price can be different from the actual spot-price at the time we conducted the experiment, since "TheCompensators*" buy batches of licenses at a specific price and then retire them based on the donations they receive, irrespective of price-changes that appear in the meantime.

suring, among others, their attributed importance to climate protection, and their belief in the effectiveness of retiring emission rights as a measure to protect the climate. In order to find out whether reactions to the different types of transparency can be explained by psychological reactance, we have two approaches. First, we assess participants' perception of the default value as freedom threatening, autonomy-decreasing, manipulative, and pressuring, as well as its tendency to evoke negative emotional reactions, such as irritation, anger, annoyance, and aggravation. We refer to this as state reactance (Dillard & Shen, 2005). Second, we measure subjects' proneness to psychological reactance, referred to as trait reactance, with Hong's Psychological Reactance Scale (Hong & Faedda, 1996). Both measures were assessed after subjects made their decision of how much to contribute.⁷ Relevant questions are in Appendix C.

4. Results

We present and discuss findings in the following way: First, we demonstrate our main results regarding the effectiveness of defaults and their interrelation with transparency. Second, we analyze the measures used to investigate the relevance of psychological reactance theory to transparency of defaults.

⁷We assume that measuring reactance items before treatments would have introduced an "additional nudge" with a potential influence on contributions. Kruskal-Wallis tests and Steel-Dwass-Critchlow-Fligner multiple comparison tests do not show any significant difference between treatments for all state and trait reactance items. This suggests there is no significant effect of treatments. However, we cannot completely exclude a potential common impact of all treatments on reactance.

4.1. *Default effects*

Overall, 214 subjects contributed 562 Euro to retire carbon licenses, resulting in 2.63 Euro per subject. The average distance of contributions to the default value was 5.54 Euro.⁸ Of all participants, 64.95% contributed a positive amount, and 11.68% opted for the default value. Table 2 presents summary statistics of the variables divided by experimental groups. Figure 1 presents the respective mean contributions.

We focus our analysis of treatment effects on two outcome-categories: contributions, and the convergence of decisions toward the default value. Within the first dimension we consider two variables, i.e. frequency of positive contributions (extensive margin) and the level of contributions. Within the second dimension we consider the distance of each contribution to the default value, and the frequency of contributions of this value. Our hypotheses specifically refer to treatment differences of aggregated contributions, but 'default-effects' can be further differentiated. By looking at these distinct outcome variables, we allow for a deeper understanding of potential treatment-effects.

We use non-parametric tests because we identified outliers, and Shapiro-Wilk normality tests reject the hypothesis that contributions, as well as distances to the default value, are normally distributed ($W = 0.806$, $p = 1.374 \times 10^{-15}$ for contributions; $W = 0.819$, $p = 4.87 \times 10^{-15}$ for distances). This is caused primarily by a large amount of zero contributions compared to other possible values (see Figure B.4 in Appendix B).

⁸Positive and negative distances to the default are treated equally. Thus, e.g., contributions of 6 Euro and 10 Euro are both interpreted as a distance of 2 Euro.

Testing H1 with a Mann-Whitney test rejects the null hypothesis of equal contributions at 5% significance between Control vs. Default ($W = 707.5$, $p = 0.007$), Control vs. Default+Purpose ($W = 642$, $p = 0.028$), and Control vs. Default+Info+Purpose ($W = 687$, $p = 0.033$). We reject H1 in case of a mere default, as well as in case of a default accompanied by its purpose for all outcome variables. For reasons of brevity we will only report statistics for tests of equal contributions, and include all remaining p-values in Tables B.7 - B.10, in Appendix B. In case of a default with information on its potential influence, the difference of contributions ($W = 769.5$, $p = 0.084$), distances to the default value, as well as the fraction of subjects contributing are only marginally significant at 10%. We do not reject the hypothesis that subjects in the Default+Info group picked the default value as often as subjects in the control group did. In case of a default accompanied by both types of transparency, differences to the control group are significant for the remaining outcome variables except for the fraction of subjects donating. By performing post-hoc observed power analyses for the pairwise Mann-Whitney tests we calculate power coefficients of 0.69, 0.26, 0.46, and 0.46, for tests comparing contributions in the Control vs. Default, Default+Information, Default+Purpose, and Default+Info+Purpose groups, respectively.

Overall, we find evidence for H1, i.e. that there is a default- and pull-effect. However, evidence is limited for a default combined with information on its potential influence on behavior.

The pull-effect becomes tangible when looking at Table 2. More than half of the participants in the control group contributed zero, with no contributions between 6 Euro and 9 Euro. In all treatment groups, the fraction of

subjects that contributed nothing is lower, and more subjects chose the default value than in the control group. Together with findings outlined above, this suggests that an overall increase of average contributions in the Default, as well as Default+Purpose groups is partially due to an increase in the fraction of subjects that contributed, as well as a higher fraction of subjects that chose the default, or values near the default. In the Default+Info group, we find a marginally significant increase in the extensive margin, whereas in the Default+Info+Purpose group we find a higher fraction of people picking the default value.

To check robustness of the default effect we focus on contributions as an outcome variable in regression. We begin with a restricted model limited to the treatment variable, then add a dummy variable indicating that subjects perceive climate protection to be (very) important, and proceed to add other relevant covariates (Table 3). The reason we add importance to protect the climate separately is that a Chi²-Test rejects the hypothesis that subjects are equally distributed among the treatment groups with respect to this variable. The test is marginally significant for Control vs. Default ($\chi^2(1) = 3.503$, $p = 0.061$), and for Control vs. Default+Info ($\chi^2(1) = 3.666$, $p = 0.056$).

By controlling for this variable we ensure that estimates of treatment effects are not conditionally biased. In theory, since covariates come from a questionnaire that is taken by subjects after being exposed to treatments, there is a risk of the respective manipulations being the reason for the differences in importance-ratings. However, we argue that, in case of endogeneity, we would expect this difference also to be significant for the remaining two treatment groups. This is not the case both for comparisons

of Control vs. Default+Purpose ($\chi^2(1) = 0.142$, $p = 0.706$) and Control vs. Default+Info+Purpose ($\chi^2(1) = 0.098$, $p = 0.754$).⁹ We therefore judge controlling for perceived importance of climate protection as an adequate mean to estimate unbiased treatment effects.

Regarding Tobit models in Table 4, un-restricted model (3) includes all covariates, i.e. rating of the importance of climate protection, gender, age, no previous experience with experiments, and judgment of buying emission licenses from the EU ETS as an ineffective tool for climate protection. The Tobit model accounts for left-censored contributions. This means we assume that at least some subjects would choose to take from instead of contribute to the public good. This assumption is common in dictator-games and empirically valid (Engel, 2011).

Model (1) predicts that a mere default, a default plus its purpose, as well as a default plus both types of transparency lead to higher average contributions compared to no default. The effect of Default+Info is only marginally significant. When controlling for subjects' perception of the importance of climate protection in model (2), all coefficients, except that of Default+Purpose, decrease. This results in non-significance for Default+Info. As argued above, we judge these coefficients to be conditionally unbiased, and therefore more appropriate in order to estimate treatment effects. The positive effect of *Importance of CP* on the outcome is highly significant. A likelihood-ratio test suggests that the un-restricted model fits the data sig-

⁹Additionally, calculating VIFs for independent variables of the full regression model indicates that multicollinearity is not an issue, with the largest VIF being lower than 10. Furthermore, a Durbin-Wu-Hausman test suggests that endogeneity of Importance of CP is not an issue.

nificantly better ($\chi^2(1) = 14.47$, $p = 1 \times 10^{-4}$). Controlling for additional covariates increases precision of the estimated average treatment effects. A likelihood-ratio test suggests that un-restricted model (3) fits the data significantly better than restricted model (2) ($\chi^2(4) = 26.34$, $p = 2.7 \times 10^{-5}$).

Based on the full model, a default, as well as a default supplemented by its purpose, increase average contributions to the public good by 2.36 Euro and 2.52 Euro, respectively, compared to no default. A default with full transparency increases contributions by 1.95 Euro.

F1: There is a default effect on contributions for a default, a default with added purpose, as well as for a default with both types of transparency.

4.2. Influence of transparency on default effectiveness

We proceed testing H2 – H4. A Kruskal-Wallis test for equal contributions in the treatment groups is insignificant ($H(3) = 1.117$, $p = 0.773$). So are respective pairwise comparisons with Dunn’s test (not reported). Contributions in the treatment groups do not significantly differ from each other. The same holds for distances to the default value ($H(3) = 1.459$, $p = 0.692$), the fraction of subjects choosing to contribute ($\chi^2(3) = 0.860$, $p = 0.835$, Chi²-Test), as well as the fraction of subjects choosing the default value ($p = 0.346$, Fisher’s exact test). Consequently, for neither of the four outcome variables we find significance in the differences suggested by H2, i.e. between Default and Default+Info; H3, i.e. between Default+Purpose and Default; and H4, i.e. between Default+Purpose, Default+Info+Purpose, and Default+Info, even though effects go into the hypothesized direction (Table 2).

Assuming the use of a one-way ANOVA (as the parametric alternative to the Kruskal-Wallis test), and an observed within-groups variance of 9.48, post-hoc performed observed power analysis results in a minimum detectable effect size of $f = 0.26$ (for a power of 0.8). However, the observed effect size of transparency on contributions is small ($f = 0.09$), again assuming a one-way ANOVA. This results in an observed power coefficient of 0.14. As above, we augment our analysis by focusing on contributions in stepwise Tobit-regression (Table 4). In un-restricted model (3), an omnibus Wald-test for equality of parameter estimates for Default, Default+Info, Default+Purpose, and Default+Info+Purpose does not allow us to reject the null hypothesis at any conventional significance level ($F(3, 205) = 1.11, p = 0.344$). The same holds for restricted models. We find no evidence of unequal contributions in the treatment groups. Consequently, we find no evidence that transparency significantly reduces contributions¹⁰.

F2: Informing participants that the default may have an influence on their decision does not significantly decrease contributions compared to when they are not informed.

F3: Informing participants about the default’s purpose does not significantly increase contributions compared to when they are not informed.

F4: Informing participants that the default may have an influence on

¹⁰We plot estimated treatment-effects of un-restricted regression models in Appendix B (Figures B.5, B.6, and B.7).

their decision, as well as of the default’s purpose does not decrease or increase contributions, compared to the other types of transparency (including no transparency at all).

Regarding the additional covariates, *Gender* and *EUETS not effective* are significant. Being male, as well as judging the EU ETS as not effective to protect the climate, decrease average contributions. The former finding is consistent with evidence from dictator games (Engel, 2011). Findings on gender differences in public good games are ambiguous, however (Croson & Gneezy, 2009). In the context of real contributions to climate protection, evidence by Diederich & Goeschl (2014), while suggesting that female subjects are less indifferent to climate protection, do not support a higher willingness to pay for emission certificates of women. Findings with respect to age somewhat align with those of Borghans & Golsteyn (2015) who find, in a less restricted sample, that the default effect does vary with age. However, at around 22 years (the mean of our sample) they find a relatively large default effect. This may explain why we find a default effect, but no effect of age.

4.3. *Psychological reactance and transparency*

In order to find out whether reactions to different types of transparency can be explained by psychological reactance, we create an index for each of the two state reactance-categories, i.e. for the perceived threat to freedom and the anger-category.¹¹

¹¹We constructed a dummy-variable, which is equal to 1 when the subject "agreed" or "strongly agreed", resp. replied with "to some extent" or "very" to the respective statements, for each item (see Appendix C). Then, we added the respective dummies in

We model the log odds of subjects being in a higher level of each of both ordinal indexes on all explanatory variables used above (Table 5). Note that this regression excludes observations from the control group since subjects in this group were not presented with the default option.

None of the coefficients modeling treatment effects are significant.¹²

F5: Combining the default with information about its potential behavioral influence does not increase participants' experience of state reactance.

We find a significant negative influence of age on experienced anger triggered by the default value. The finding that experiencing negative emotions decreases with age is known in the literature (e.g. Charles et al., 2001).

To further test if reactions towards the combination of a default value with different types of transparency can be explained by psychological reactance, we measured subjects' proneness to experience psychological reactance. To create an index for trait reactance, we proceeded similarly to the case of state reactance.¹³

Specifically, we test whether subjects' reactions towards different types of transparency accompanying the default differ depending on subjects' trait re-

each category, to form two indexes, each ranging from zero to four. Findings are consistent for when both dependent variables are included as (un-weighted) factor-based scores in linear OLS-regression.

¹²This finding is consistent with non-parametric tests for differences of individual items of the scales (not reported).

¹³We constructed dummy variables for each of the 14 items of the scale, which are equal to 1 when the subject responded with "Agree" or "Strongly agree" to the respective question, 0 otherwise. We then added the dummies for each subject to create the index, which ranges from zero to 14. Findings are consistent for trait reactance included as a (un-weighted) factor-based score.

actance. Therefore, we run regressions with an interaction term of treatment variable and trait reactance index. The latter is centered on the mean, so that treatment-main-effects are meaningful (Table 6). Note that this regression excludes observations from the control group. For reasons of brevity, we focus on the main effects of trait reactance, as well as on interaction-effects.

As in previous Tobit models, model (7) fits the data better than model (6) ($\chi^2(1) = 11.47$, $p = 7 \times 10^{-4}$), and model (8) fits the data better than model (7) ($\chi^2(4) = 19.71$, $p = 6 \times 10^{-4}$). We find no significant main effect of trait reactance, nor do we find that the different types of transparency and the trait reactance index interact significantly for any of the three model-specifications. In other words, there is no evidence that the effect of different types of transparency on average contributions is conditional on subjects' trait reactance.

F6: The influence of information on the default effect does not depend on the level of trait reactance of participants.

Both approaches that are linking different types of transparency of a default to psychological reactance suggest that subjects neither perceive a default value differently based on the type of transparency accompanying it, nor does their inherent propensity to show psychological reactance change the way they react to these different types of transparency.

5. Discussion and conclusion

This study contributes to the discussion of nudges and transparency by providing empirical evidence. Despite the widespread application of nudges, some people are concerned of the potentially manipulative nature of behavioral interventions. In democratic societies, public authorities are expected to be transparent with regard to their actions and intentions. Therefore, covertly 'exploiting' people's psychological biases might have an impact on perceived legitimacy, and ultimately effectiveness of such policies. The most straightforward solution to this problem is to instruct policy-makers to disclose information regarding the potential influence of the nudge, and its purpose. However, this suggestion raises the concern that nudges will no longer be effective. As expressed by Bovens (2009), nudges "work best in the dark". The results of this study suggest that this concern might be overstated.

Our findings suggest that a default value is an effective way of increasing individual voluntary contributions to climate protection. Increased aggregate contributions are consistent with inertia, as well as anchoring. A higher fraction of participants picking the default value instead of specifying another amount in the default, compared to the control group, supports the inertia/ effort reduction explanation. However, deviation costs in the experiment are marginal (the subject had to make two mouse-clicks, as well as to type in the contribution amount, instead of just making one mouse click on the default button), and contributing the default value is also consistent with an anchoring explanation: Subjects may choose the default value not only because of inertia, but also because they consider this value first and only then employ reasons against this value, conditional on what they wanted

to contribute initially. This anchoring-explanation is consistent both with picking the default or moving towards the default, whereas inertia is only consistent with the former behavior (Dhingra et al., 2012). We observe that subjects who contribute a positive amount do contribute more on average, when there is a default value with either type of transparency, but the differences to the control group are not significant. Additionally, we observe an increase of subjects giving a positive amount due to the default, which is consistent with the anchoring explanation. Together, our findings suggest that increased aggregate contributions in the default groups are due to an increase of the fraction of subjects contributing, as well as of an increase of the fraction of subjects choosing the default value, but not because of increased average contributions of subjects that contribute. Inertia, as well as anchoring may therefore both be reasons for why we observe default effects. Intuitively, we would expect anchoring to play a more pronounced role in real world applications of pro-environmental nudges, especially if defaults result in repeated and/or significant financial costs. For someone who highly values environmental- and climate protection, deviating from a default, which may be perceived as conveying information about social norms, can incur non-financial costs, especially if he or she aims to uphold a positive self-image. Maintaining a positive self image, as well as being consistent with social norms, can be achieved by decreasing (not necessarily closing) the gap between default value and initially intended contribution.

Our experiment provides evidence that defaults increase contributions to climate protection even when complemented by a disclosure regarding the potential influence of the default and its purpose, or just its purpose.

However, it is not clear whether informing subjects only of the potential influence of the default has an impact on contributions. Furthermore, we find no evidence that additional information on the potential behavioral influence and/or purpose of the default triggers psychological reactance. Likewise, we find no evidence that subjects differing in their proneness to experience reactance also differ in how they react towards the default with additional information.

These findings suggest that despite the initial concern over the inhibiting influence of transparency, nudges in the form of defaults can be transparent and at the same time effective. However, the type of transparency might matter. In order to preserve the effect of defaults and increase the legitimacy of behaviorally informed policies, policy makers should be transparent about their motives. The motive itself and how it is perceived by the decision maker has been found to matter for advice (Kuang et al., 2007), and might also matter for defaults, as well as nudges in general.

Our findings support and add to previous evidence on the influence of transparency. Loewenstein et al. (2015) and Kroese et al. (2016) found that pro-self defaults were effective in health contexts even after disclosing information about them. Our study extends this conclusion to pro-social nudges, a type that is widely used in the context of public policy-making. Moreover, we extend findings by Steffel et al. (2016) by examining the influence of transparency in a more realistic setting where participants' decisions have an actual consequence for them, and for the environment. Findings are also useful for the private sector and NGOs aiming to include nudges in their inventory to increase contributions to environmental protection, and possibly

other public goods, e.g. charity.

Although several recent studies link nudges to psychological reactance, they do so either indirectly, or they deal with hypothetical and attitudinal, instead of behavioral outcomes (Haggag & Paci, 2014; Arad & Rubinstein, 2015; Loewenstein et al., 2015; Hedlin & Sunstein, 2016). By measuring both state and trait reactance, we enable a more direct way of assessing the role of psychological reactance in the influence of transparency on the effectiveness of a default value. To our best knowledge, Goswami & Urminsky (2016) is the only study that assesses the interaction of trait reactance with the size of a default value on behavioral outcomes, i.e. charitable giving. They find no significant interaction effect. On a more general level, our findings, in line with theirs, suggest that psychological reactance plays a lesser or no role with respect to behavioral effects of defaults, and, in our case, transparency.

Further research could evaluate the role of trait reactance on how subjects respond to different types of transparency for different types of nudges, i.e. social norms or framing. Since our experiment has a rather limited amount of subjects, field experiments can establish statistically more powerful findings for interaction effects. Due to a more realistic context, a field experimental approach would also increase external validity. Nevertheless, our experiment is less abstract than a 'regular' laboratory experiment due to the fact that contributions have a real effect on climate protection Harrison & List (2004). The current study focuses on one type of nudge, and a specific context. Further research is needed in order to determine the overall influence of transparency on the effectiveness of nudges. Moreover, results might be context-specific, thus requiring further investigation into pro-social

nudges. Delving into the welfare implications of transparency can also become a promising research endeavor (Sunstein, 2015).

Overall, our findings advance the understanding of how nudges in general, and defaults specifically, affect individual behavior with social consequences, and how policy-makers can increase their transparency without limiting their effectiveness.

References

- Alemanno, A., & Sibony, A.-L. (2015). *Nudge and the law: A European perspective*. Oxford and Portland, OR: Bloomsbury Publishing.
- Allcott, H., & Mullainathan, S. (2010). Energy behavior and energy policy. *Science*, *327*, 1204–1205.
- Arad, A., & Rubinstein, A. (2015). The people’s perspective on libertarian–paternalistic policies.
- Araña, J. E., & León, C. J. (2013). Can defaults save the climate? evidence from a field experiment on carbon offsetting programs. *Environmental & Resource Economics*, *54*, 613–626.
- Bartling, B., Fehr, E., & Herz, H. (2014). The intrinsic value of decision rights. *Econometrica*, *82*, 2005–2039.
- Bolton, G. E., & Ockenfels, A. (2012). Behavioral economic engineering. *Journal of Economic Psychology*, *33*, 665–676.
- Bordalo, P., Gennaioli, N., & Shleifer, A. (2012). Salience theory of choice under risk. *The Quarterly Journal of Economics*, *127*, 1243–1285.

- Borghans, L., & Golsteyn, B. H. (2015). Susceptibility to default training options across the population. *Journal of Economic Behavior & Organization*, 117, 369–379.
- Bovens, L. (2009). The ethics of nudge. In T. Grüne-Yanoff, & S. O. Hansson (Eds.), *Preference change: Approaches from philosophy, economics and psychology* (pp. 207–219). Dordrecht: Springer Science & Business Media.
- Brehm, J. W. (1966). *A theory of psychological reactance*. Oxford, England: Academic Press.
- Brehm, S. S., & Brehm, J. W. (2013). *Psychological reactance: A theory of freedom and control*. New York: Academic Press.
- Cappelletti, D., Mittone, L., & Ploner, M. (2014). Are default contributions sticky? an experimental analysis of defaults in public goods provision. *Journal of Economic Behavior & Organization*, 108, 331–342.
- Charles, S. T., Reynolds, C. A., & Gatz, M. (2001). Age-related differences and change in positive and negative affect over 23 years. *Journal of Personality and Social Psychology*, 80, 136–151.
- Chartrand, T. L., Dalton, A. N., & Fitzsimons, G. J. (2007). Nonconscious relationship reactance: When significant others prime opposing goals. *Journal of Experimental Social Psychology*, 43, 719–726.
- Croson, R., & Gneezy, U. (2009). Gender differences in preferences. *Journal of Economic Literature*, 47, 448–474.

- Dhingra, N., Gorn, Z., Kener, A., & Dana, J. (2012). The default pull: An experimental demonstration of subtle default effects on preferences. *Judgment and Decision Making*, 7, 69–76.
- Diederich, J., & Goeschl, T. (2014). Willingness to pay for voluntary climate action and its determinants: Field-experimental evidence. *Environmental & Resource Economics*, 57, 405–429.
- Dillard, J. P., & Shen, L. (2005). On the nature of reactance and its role in persuasive health communication. *Communication Monographs*, 72, 144–168.
- Dolan, P., Hallsworth, M., Halpern, D., King, D., Metcalfe, R. D., & Vlaev, I. (2012). Influencing behaviour: The mindspace way. *Journal of Economic Psychology*, 33, 264–277.
- Engel, C. (2011). Dictator games: a meta study. *Experimental Economics*, 14, 583–610.
- Fehr, E., Herz, H., & Wilkening, T. (2013). The lure of authority: Motivation and incentive effects of power. *The American Economic Review*, 103, 1325–1359.
- Felsen, G., Castelo, N., & Reiner, P. B. (2013). Decisional enhancement and autonomy: Public attitudes towards overt and covert nudges. *Judgment and Decision Making*, 8, 202–213.
- Fischbacher, U. (2007). z-tree: Zurich toolbox for ready-made economic experiments. *Experimental Economics*, 10, 171–178.

- Goswami, I., & Urminsky, O. (2016). When should the ask be a nudge? the effect of default amounts on charitable donations. *Journal of Marketing Research*, 53, 829–846.
- Haggag, K., & Paci, G. (2014). Default tips. *American Economic Journal: Applied Economics*, 6, 1–19.
- Hagman, W., Andersson, D., Västfjäll, D., & Tinghög, G. (2015). Public views on policies involving nudges. *Review of Philosophy and Psychology*, 6, 439–453.
- Hansen, P. G., & Jespersen, A. M. (2013). Nudge and the manipulation of choice: A framework for the responsible use of the nudge approach to behaviour change in public policy. *European Journal of Risk Regulation*, 4, 3–28.
- Harrison, G. W., & List, J. A. (2004). Field experiments. *Journal of Economic Literature*, 42, 1009–1055.
- Hausman, D. M., & Welch, B. (2010). Debate: To nudge or not to nudge. *Journal of Political Philosophy*, 18, 123–136.
- Hedlin, S., & Sunstein, C. R. (2016). Does active choosing promote green energy use? experimental evidence. *Ecology Law Quarterly*, 43, 107–142.
- Hong, S.-M., & Faedda, S. (1996). Refinement of the hong psychological reactance scale. *Educational and Psychological Measurement*, 56, 173–182.
- House of Lords Report (2011). Behaviour change.

- Kroese, F. M., Marchiori, D. R., & Ridder, D. T. D. d. (2016). Nudging healthy food choices: a field experiment at the train station. *Journal of Public Health*, *38*, e133–e137.
- Kuang, X., Weber, R. A., & Dana, J. (2007). How effective is advice from interested parties? an experimental test using a pure coordination game. *Journal of Economic Behavior & Organization*, *62*, 591–604.
- Loewenstein, G., Bryce, C., Hagmann, D., & Rajpal, S. (2015). Warning: You are about to be nudged. *Behavioral Science & Policy*, *1*, 35–42.
- Lourenco, J. S., Ciriolo, E., Almeida, S. R., & Troussard, X. (2016). Behavioural insights applied to policy: European report 2016.
- Owens, D., Grossman, Z., & Fackler, R. (2014). The control premium: A preference for payoff autonomy. *American Economic Journal: Microeconomics*, *6*, 138–161.
- Perino, G. (2015). Climate campaigns, cap and trade, and carbon leakage: Why trying to reduce your carbon footprint can harm the climate. *Journal of the Association of Environmental and Resource Economists*, *2*, 469–495.
- Rebonato, R. (2014). A critical assessment of libertarian paternalism. *Journal of Consumer Policy*, *37*, 357–396.
- Reisch, L. A., & Sunstein, C. R. (2016). Do europeans like nudges? *Judgment and Decision Making*, *11*, 310–325.
- Steffel, M., Williams, E. F., & Pogacar, R. (2016). Ethically deployed de-

- faults: Transparency and consumer protection through disclosure and preference articulation. *Journal of Marketing Research*, 53, 865–880.
- Sunstein, C. R. (2014a). *Simpler: The future of government*. New York: Simon and Schuster.
- Sunstein, C. R. (2014b). *Why nudge? The politics of Libertarian Paternalism*. New Haven, CN: Yale University Press.
- Sunstein, C. R. (2015). Fifty shades of manipulation. *Journal of Behavioral Marketing*, in press, .
- Sunstein, C. R. (2016). Do people like nudges? *Administrative Law Review*, in press, .
- Sunstein, C. R., & Reisch, L. A. (2016). Behaviorally green: Why, which and when defaults can help. In F. Beckenbach, & W. Kahlenborn (Eds.), *New Perspectives for Environmental Policies through Behavioral Economics* (pp. 161–194). Heidelberg, New York, Dordrecht, London: Springer.
- Thaler, R. H., & Sunstein, C. R. (2008). *Nudge: Improving decisions about health, wealth, and happiness*. New Haven, CN: Yale University Press.
- World Bank (2015). World development report 2015: Mind, society, and behavior.

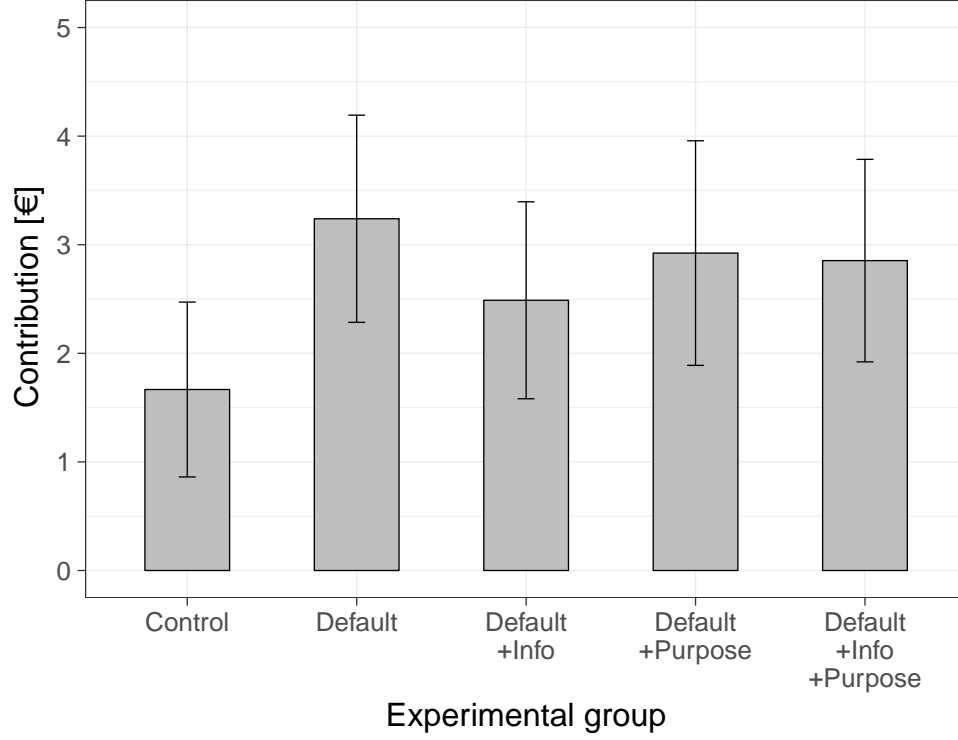
Table 1: Experimental design

Experimental group	Default value	Transparency information
Control	No	No information
Default	8 Euro	No information
Default+Info	8 Euro	"Please consider that the preselected default value might have an influence on your decision."
Default+Purpose	8 Euro	"Please consider that the preselected default value is meant to encourage higher contributions for the climate protection fund."
Default+Info+Purpose	8 Euro	"Please consider that the preselected default value might have an influence on your decision. This is meant to encourage higher contributions for the climate protection fund."

Table 2: Descriptive statistics of all outcome variables to assess the default effect

Experimental group	Variable and respective statistic						
	Contri- bution		Distance		Con- tributed	Picked default	n
	Mean	SD	Mean	SD	Mean	Mean	
Control	1.67	2.68	6.6	1.91	46.67	0	45
Default	3.24	3.21	4.89	3.01	73.91	19.57	46
Default+Info	2.49	2.95	5.74	2.45	67.44	6.98	43
Default+Purpose	2.92	3.19	5.28	2.83	71.79	15.38	39
Default+Info+Purpose	2.85	2.95	5.15	2.95	65.85	17.07	41

Figure 1: Mean contributions per experimental group



Notes: Error bars represent 95% confidence intervals.

Table 3: Descriptive statistics of covariates

Experimental group	Variable and respective statistic					
	Age		Gender (Male)	Importance of CP	No exp. Exp- erience	EUETS not effective
	Mean	SD	Mean	Mean	Mean	Mean
Control	21.8	3.1	60	57.8	31.1	57.8
Default	22	2.8	60.9	78.3	30.4	60.9
Default+Info	22	3	51.2	79.1	20.9	53.5
Default+Purpose	22.3	4.7	53.8	63.4	19.5	58.5
Default+Info+Purpose	22.7	3.7	58.5	51.3	20.5	64.1

Table 4: Stepwise Tobit-Regression

	(1) Con.	(2) Con.	(3) Con.
Default	2.706** (0.962)	2.229* (0.938)	2.363** (0.892)
Default+Info	1.703 ^x (0.983)	1.233 (0.977)	1.065 (0.898)
Default+Purpose	2.307* (1.007)	2.510* (0.984)	2.518** (0.910)
Default+Info+Purpose	2.058* (0.996)	1.863 ^x (0.951)	1.946* (0.913)
Importance of CP		2.485*** (0.673)	2.166** (0.660)
Gender (Male=1)			-1.659** (0.551)
Age			-0.0518 (0.0816)
No exp. Experience			-0.345 (0.673)
EUETS not effective			-2.400*** (0.569)
Constant	-0.129 (0.737)	-1.595 ^x (0.824)	2.358 (1.965)
Sigma			
Constant	4.160*** (0.250)	4.020*** (0.254)	3.777*** (0.229)
Observations	214	214	214
Log Pseudolikelihood	-456.93	-449.69	-436.52
F	(4, 210)=2.21	(5, 209)=4.64	(9, 205)=6.57
Prob > F	0.0689	0.0005	0.0000
Pseudo R ²	0.0103	0.026	0.0545

Robust standard errors in parentheses

^x $p < 0.10$, * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Table 5: Ordered logistic model of state reactance

	(4) ThreatToFreedom	(5) Anger
Default+Info	0.0933 (0.401)	-0.342 (0.395)
Default+Purpose	-0.00976 (0.456)	-0.0155 (0.454)
Default+Info+Purpose	-0.0443 (0.357)	-0.588 (0.478)
Importance of CP	-0.129 (0.307)	-0.349 (0.352)
Gender (Male=1)	-0.142 (0.303)	-0.199 (0.328)
Age	-0.0368 (0.0373)	-0.0912* (0.0422)
No ex. Experience	0.00944 (0.303)	0.0802 (0.420)
EUETS not effective	-0.0180 (0.321)	0.0365 (0.331)
Cut1		
Constant	-2.703** (0.950)	-2.098 ^x (1.111)
Cut2		
Constant	-1.918* (0.917)	-1.478 (1.126)
Cut3		
Constant	-0.729 (0.890)	-0.712 (1.133)
Cut4		
Constant	0.523 (0.875)	-0.123 (1.156)
Observations	169	169
Log Pseudolikelihood	-264.318	-199.502
Wald Chi ² (8)	1.61	9.21
Prob > Chi ²	0.9907	0.3246
Pseudo R ²	0.0029	0.0203

Robust standard errors in parentheses

^x $p < 0.10$, * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Table 6: Stepwise Tobit-models with interaction term

	(6) Con.	(7) Con.	(8) Con.
Default+Info	-0.914 (0.880)	-0.914 (0.868)	-1.217 (0.832)
Default+Purpose	-0.448 (0.902)	0.204 (0.916)	0.115 (0.889)
Default+Info+Purpose	-0.648 (0.907)	-0.381 (0.849)	-0.462 (0.858)
Reactance	-0.0743 (0.220)	-0.119 (0.211)	-0.0397 (0.210)
Default+Info \times Reactance	-0.249 (0.276)	-0.219 (0.269)	-0.193 (0.254)
Default+Purpose \times Reactance	0.166 (0.338)	0.227 (0.340)	0.0687 (0.316)
Default+Info+Purpose \times Reactance	0.0502 (0.298)	0.0597 (0.279)	-0.111 (0.278)
Importance of CP		2.431** (0.741)	2.120** (0.718)
Gender (Male=1)			-1.597* (0.648)
Age			0.0231 (0.0775)
No exp. Experience			-0.485 (0.735)
EUETS not effective			-2.258*** (0.593)
Constant	2.615*** (0.616)	0.721 (0.844)	3.112 ^x (1.846)
Sigma			
Constant	4.022*** (0.244)	3.890*** (0.246)	3.686*** (0.232)
Observations	169	169	169
Log Pseudolikelihood	-376.92	-371.19	-361.33
F	(7, 162)=0.83	(8, 161)=2.15	(12, 157)=3.84
Prob > F	0.5622	0.0337	0.0000
Pseudo R ²	0.0053	0.0204	0.0464

Robust standard errors in parentheses

^x $p < 0.10$, * $p < 0.05$, ** $p < 0.01$, *** $p < 0.001$

Appendix A. Experimental design

Instructions

Welcome and thank you very much for participating in this experiment. This experiment is about decision-making. Please read the following instructions carefully. Everything that you need to know in order to participate in this experiment is explained below. If you have any difficulties in understanding these instructions please raise your hand and I will come to you. Please note that communication between participants is strictly prohibited during the experiment. Communication between participants will lead to the exclusion from the experiment. The experimental procedure will be as follows. You will receive 10 Euro. Please decide how much of the 10 Euro you would like to spend on climate protection. You can choose freely how much, if any, you contribute to climate protection (whole numbers between 0-10). Should you decide to contribute, we will realize your contribution to climate protection by buying and retiring carbon emission licenses from the European Union Emissions Trading System (EU ETS) at the end of the experiment (please read the respective paragraph below for a description). By this, you have the possibility to make a real contribution to climate protection. The rest of the money is your private pay-out that you will receive in cash at the end of the experiment.

After making the decision you will be kindly asked to complete a short questionnaire. Please note that your decisions in this experiment are anonymous and will not be revealed at any stage to the other participants. (If relevant) a confirmation of the aggregated real payment to the climate protection fund will be sent to all participants at the end of the whole experiment.

The Climate Protection Fund

If a person wants to protect the climate, emitting climate gases such as CO₂ should be avoided. But it is possible to do even more: Individuals can buy and delete emission certificates from the EU Emission Trading System (ETS) through certified organizations and NGOs. By doing so, a private person reduces the amount of CO₂ which can be emitted by European industries, protects the environment and ensures that the development of climate-friendly technologies is accelerated. In this experiment, the participants' contributions to the climate protection fund will be used to buy real carbon dioxide (CO₂) emission licenses on the market of the European Union Emissions Trading Scheme (EU ETS) via the website "TheCompensators.org". It is one example of an NGO that allows ordinary people to directly participate in the EU ETS scheme, and where they can make decisions on CO₂ reductions.

The following table shows how much kilograms of carbon you reduce with your payment, and how much money you receive for yourself. The far right row indicates the respective amount of reduced CO₂ relative to a Dutch citizens' average of 9163 kg of CO₂ emitted per year.

For example, with a payment of 3 Euro to retire carbon licenses, you retire 542 kg CO₂. This corresponds to approximately 6% of the average emissions per capita per year of a Dutch person. As a private pay-out you get 7 Euro. With a payment of 8 Euro to retire carbon licenses, you retire 1,447 kg CO₂. This corresponds to approximately 16% of the average emissions per capita per year of a Dutch person. As a private pay-out you get 2 Euro.

Payment to retire CO ₂ -allowances	Private payout €	CO ₂ abated [kg]	Share of average emissions per year per person [%]
0	10 €	0	0%
1	9 €	181	2%
2	8 €	362	4%
3	7 €	542	6%
4	6 €	723	8%
5	5 €	904	10%
6	4 €	1,085	12%
7	3 €	1,266	14%
8	2 €	1,447	16%
9	1 €	1,627	18%
10	0 €	1,808	20%

Figure A.2: Experimental screen for Control

Periode

1 von 1

Verbleibende Zeit (sec): 44

You are given 10 Euros.

Please decide how much of your 10 Euros you would like to allocate to the climate protection fund.

My contribution to the climate protection fund:

OK

Figure A.3: Experimental screen for Default + transparency

Periode

1 von 1

Verbleibende Zeit (sec): 53

You are given 10 Euros.

Please decide how much of your 10 Euros you would like to allocate to the climate protection fund.

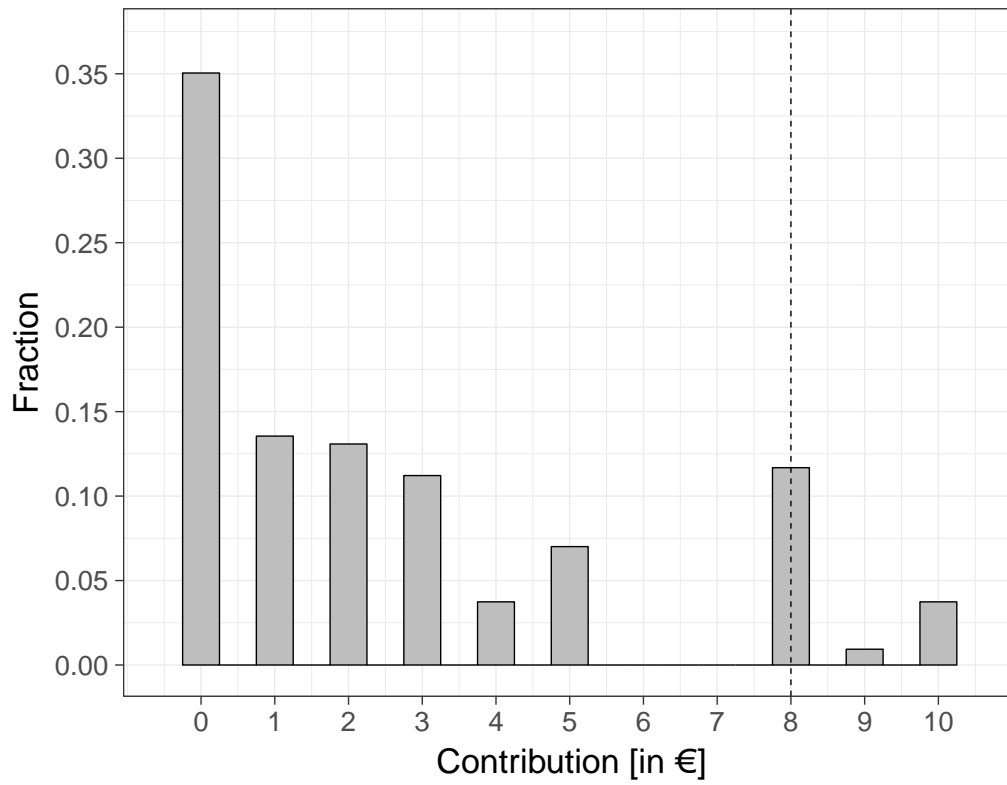
My contribution to the climate protection fund:

I would like to choose a different amount:

Respective transparency information
is shown here

Appendix B. Statistical analyses

Figure B.4: Distribution of contributions



Notes: The dashed line indicates the default value

Table B.7: P-values for pairwise MW tests of Contribution

	Control	Default	Default +Info	Default +Purpose
Default	0.007			
Default+Info	<i>0.084</i>	0.302		
Default+Purpose	0.028	0.625	0.635	
Default+Info+Purpose	0.033	0.544	0.637	0.91

Notes: $p < 0.05$ in bold, $p < 0.1$ in cursive

Table B.8: P-values for pairwise MW tests of Distance

	Control	Default	Default +Info	Default +Purpose
Default	0.003			
Default+Info	<i>0.067</i>	0.224		
Default+Purpose	0.018	0.575	0.569	
Default+Info+Purpose	0.02	0.627	0.507	0.988

Notes: $p < 0.05$ in bold, $p < 0.1$ in cursive

Table B.9: P-values for pairwise Chi²-tests of Contributed

	Control	Default	Default +Info	Default +Purpose
Default	0.015			
Default+Info	<i>0.08</i>	0.662		
Default+Purpose	0.035	1	0.851	
Default+Info+Purpose	0.116	0.558	1	0.74

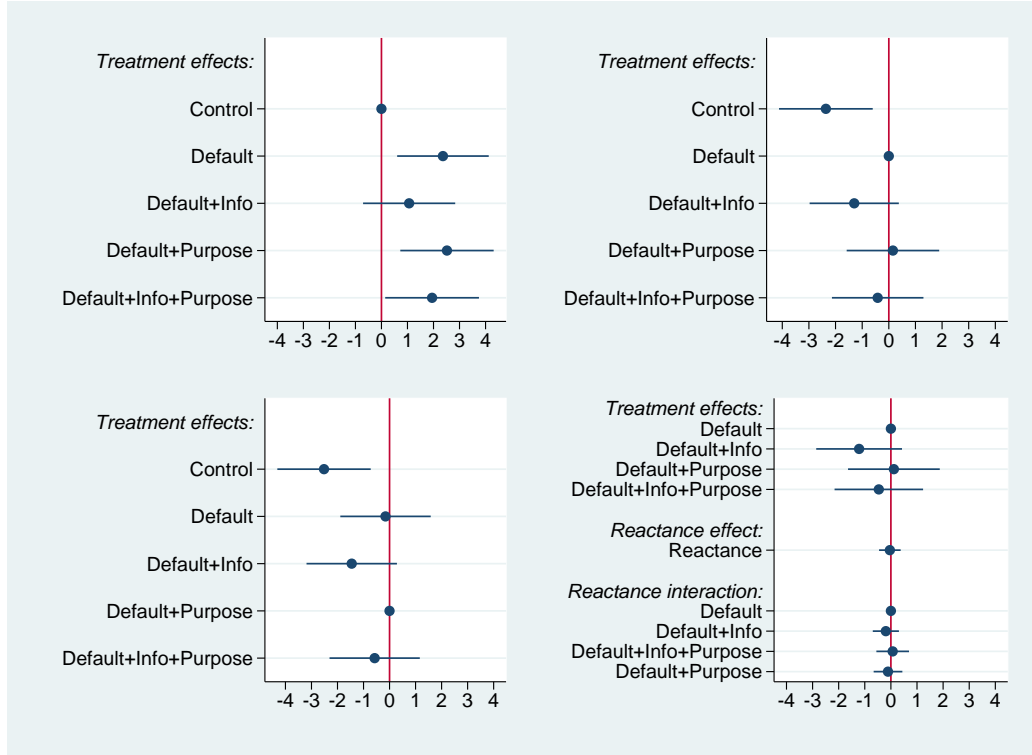
Notes: Chi²-Test with Yates continuity correction. $p < 0.05$ in bold, $p < 0.1$ in cursive

Table B.10: P-values for pairwise Fisher exact-tests of Picked default

	Control	Default	Default +Info	Default +Purpose
Default	0.003			
Default+Info	0.113	0.121		
Default+Purpose	0.008	0.777	0.297	
Default+Info+Purpose	0.004	0.79	0.19	1

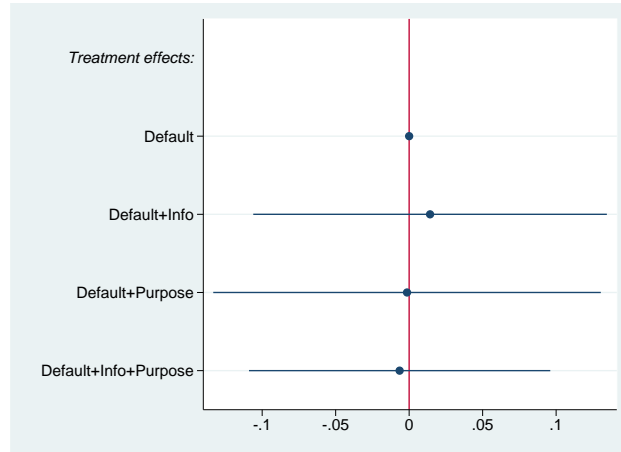
Notes: Fishers exact test for count data. $p < 0.05$ in bold, $p < 0.1$ in cursive

Figure B.5: Default and transparency effects on contributions for different base-categories



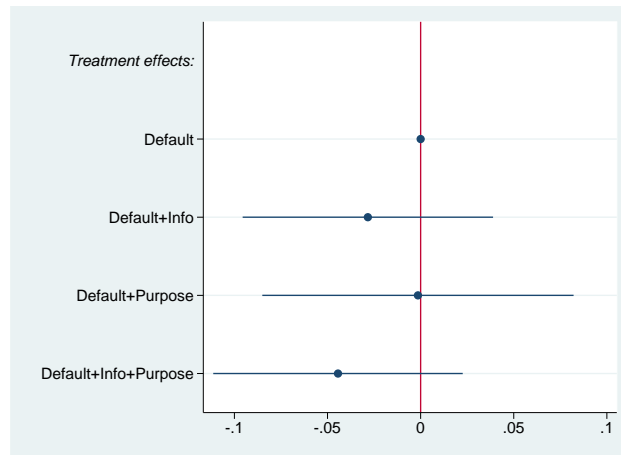
Notes: Dots with horizontal lines indicate point estimates with 95% confidence intervals from Tobit models. Dots on the zero line denote the reference category. Models (3) and (8) in Tables 4 and 6 display the underlying regression results. The top left panel refers to finding F1, the top right panel to F2 and F3, the bottom left panel to F4, and the panel on the bottom right to F6. Covariates are not shown.

Figure B.6: Default and transparency effects on perceived Threat to freedom



Notes: Dots with horizontal lines indicate point estimates with 95% confidence intervals from marginal effects of ordered logistic models. Dots on the zero line denote the reference category. Model (4) in Table 5 displays the underlying regression results (albeit not showing marginal effects). It refers to finding F5. Covariates are not shown.

Figure B.7: Default and transparency effects on Anger



Notes: Dots with horizontal lines indicate point estimates with 95% confidence intervals from marginal effects of ordered logistic models. Dots on the zero line denote the reference category. Model (5) in Table 5 displays the underlying regression results (albeit not showing marginal effects). It refers to finding F5. Covariates are not shown.

Appendix C. Questionnaire

Questionnaire on covariates

What is your gender? ☐ Male ☐ Female

What is your age?

Have you participated in other experiments before today? ☐ Yes ☐ No

How important is climate protection for you? Please circle the most suitable answer.

☐ Not important at all ☐ Not important ☐ Indifferent ☐ Important ☐ Very important

Do you think that buying real carbon dioxide (CO₂) emissions licenses on the market of the European Union Emissions Trading Scheme (EU ETS) is an effective method to contribute to climate protection? ☐ Yes ☐ No

Questionnaire on state reactance

Please indicate to what extent do you agree with the following statements on a 5-point response scale that ranges from the statement "strongly disagree" to the statement "strongly agree". (Perceived threat to freedom)

- The default value threatened my freedom to choose.

- The default value tried to make a decision for me.
- The default value tried to manipulate me.
- The default value tried to pressure me.

Please indicate to what extent do you agree with the following statements on a 5-point response scale that ranges from the statement "Not at all" to the statement "Very". (anger)

- Please indicate how irritated you were with regard to the given default value.
- Please indicate how angry you were with regard to the given default value.
- Please indicate how annoyed you were with regard to the given default value.
- Please indicate how aggravated you were with regard to the given default value.

Questionnaire on trait reactance

Please indicate to what extent do you agree with the following statements on a 5-point response scale that ranges from the statement "strongly disagree" to the statement "strongly agree".

- Regulations trigger a sense of resistance in me.
- I find contradicting others stimulating.
- When something is prohibited, I usually think, "that's exactly what I am going to do".
- The thought of being dependent on others aggravates me.
- I consider advice from others to be an intrusion.
- I become frustrated when I am unable to make free and independent decisions.
- It irritates me when someone points out things, which are obvious to me.
- I become angry when my freedom of choice is restricted.
- Advice and recommendations usually induce me to do just the opposite.
- I am content only when I am acting on my own free will.
- I resist the attempts of others to influence me.
- It makes me angry when another person is held up as a role model for me to follow.
- When someone forces me to do something, I feel like doing the opposite.
- It disappoints me to see others submitting to standards and rules.