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# Motivational Goal Bracketing: An Experiment

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## Abstract

We study theoretically and experimentally how the bracketing of non-binding goals in a repeated task affects the level of goals that people set for themselves, the actual effort provided, and the pattern of effort over time. In our model, a sophisticated or partially naïve individual sets either daily or weekly goals to overcome a motivational problem caused by present-biased preferences. In an online, real-effort experiment, we randomly allocated subjects to treatments where they set either daily goals for how much to work over a one-week period or a single weekly goal. Consistent with the theoretical predictions, in the treatment with daily goals, the aggregate goal level for the week was higher and subjects provided more effort compared to the treatment with a weekly goal. The higher effort was driven by the higher aggregate goal level. Additional treatments complemented internal commitment through goals with an externally enforced minimum work requirement to get started working each day.

**JEL Classification:** D03, D81, D91

**Keywords:** Self-control, goals, narrow bracketing, commitment, real effort, online experiment

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# 1 Introduction

Every day people face motivational problems in repeated tasks such as working, studying, dieting, exercising, or saving. While decades of research in psychology document that goals play an important role in helping people to overcome their motivational problems (e.g. Locke and Latham 2015), it is still poorly understood how goals work in repeated tasks. In such settings, a person may focus on single instances of the task and evaluate tasks relative to narrowly bracketed goals or, instead, evaluate the aggregate performance over a longer time period relative to a broadly bracketed goal. How goals are bracketed can often be linked to the way feedback about performance is given (e.g. Asch, 1990; Cadena et al., 2011), the availability of salient reference points (e.g. Pope and Schweitzer, 2011; Allen et al., 2016), or explicit advice about how to set goals.<sup>1</sup> But how does the bracketing of goals affect the level of goals that people set for themselves, the actual effort provided, and the pattern of effort over time? In this paper, we study these open questions theoretically and experimentally.

We develop a model where a sophisticated or partially naïve individual works repeatedly on a task. He sets non-binding goals to overcome the motivational problems caused by present-biased preferences. Goals can either be narrow (daily goals) or broad (a weekly goal). Our model predicts that the aggregated daily goals are higher than the weekly goal. As a consequence, individuals with daily goals work more than those with the weekly goal. The reason is the following. A weekly goal tempts individuals to put in low effort at the beginning of the week and to compensate with higher effort later (*effort substitution*). This asymmetric effort profile is suboptimal (from an ex ante perspective). Because effort costs are convex, the individual would prefer a constant effort pattern. Under plausible assumptions, a reduction in the weekly goal level leads to less variation in effort over time. Thus, taking into account evidence on the distribution of present bias, our model predicts that on average individuals adopt a lower weekly goal compared to the aggregated goal level that they would chose with daily goals.

We tested the predictions of the model using an online, real-effort experiment. Subjects

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<sup>1</sup>For example, the UK National Health Service advises daily calorie targets and weekly weighing (<https://www.nhs.uk/Tools/Pages/Losing-weight.aspx>, accessed June 2019), and 150 minutes of exercise per week (<https://www.nhs.uk/live-well/exercise/>, accessed June 2019).

were randomly assigned to set either non-binding daily goals for how much to work online in the following week or a non-binding weekly goal. The experiment mimicked a typical work-leisure self-control problem. Work was desirable (the piece-rate pay was generous) but involved unpleasant effort. Subjects faced the usual real-life temptations because our study neither required them to show up at a lab nor to obey a particular schedule.

By exogenously varying the goal bracket, we provide a clean test of the *motivational bracketing* hypothesis that narrowly bracketed goals help individuals to address their self-control problems (e.g. Read et al., 1999). It was first suggested as an explanation for why individuals who can choose their working hours, such as taxi drivers, often appear to have daily income targets (e.g. Camerer et al. 1997, Dupas and Robinson 2016). While there is much suggestive evidence for this hypothesis, clean evidence is missing.

We find support for the motivational bracketing hypothesis. Our data reveal that subjects with daily goals set a higher aggregated goal for the week than subjects with a weekly goal, and that they worked more than those with a weekly goal. The latter effect largely disappears when we control for the goal level. That is, in line with the theory, the higher effort with daily goals seems to be related to the higher goals that subjects set under the daily bracket and not to the daily bracket per se.

We extended our two baseline treatments in two directions. First, we conducted treatments that manipulated whether goals were framed as daily goals or a weekly goal, but that left the overall goal level unaffected by the framing.<sup>2</sup> These treatments allow us to directly test the prediction that the higher effort with daily goals is driven by the higher aggregated goal level compared to a weekly goal and not the framing of the goal. We confirm the prediction. Further, we test and confirm the prediction that subjects with a weekly goal work less at the beginning of the week than subjects with daily goals who face the same aggregated goal level, and that they work more at the end of the week to make up for the shortfall.

Second, to examine whether the positive effect of daily goals stems from their ability to get people started working each day, we ran additional treatments which complemented goals with a minimum work requirement. To receive any payment, subjects had to complete at

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<sup>2</sup>This treatment manipulation also is of independent interest because it allows us to test predictions of our model that also apply to settings where the goal level is unaffected by the way that goals are bracketed – for example, because the level is externally determined by an employer.

least one real-effort task per day, which took less than a minute. An innovative feature of the requirement is that it forced subjects to ‘get started’ working each day, but otherwise gave full flexibility of how to allocate work and how much to work. In the treatments with the work requirement, effort and goal levels no longer differed across treatments where subjects set daily goals or a weekly goal, and effort patterns were similar. This result is suggestive for the interpretation that forcing subjects to ‘get started’ each day mitigates problems of effort substitution with a weekly goal.

With these treatments we make another novel contribution by addressing the question of how externally enforced work requirements interact with internal commitment through goals. Surprisingly, subjects worked less if they were forced to ‘get started’ working each day (in addition to setting daily goals) than if they just set daily goals. This result can be explained by a large fraction of subjects dropping out when they were forced to work each day. Focusing only on those subjects who did not drop out, performance did not differ across treatments with and without the work requirement. This is consistent with the interpretation that daily goals on their own already get people started. The pattern is reversed for the treatments with a weekly goal. For subjects who did not drop out, effort was significantly higher with the requirement than without it - as one would expect if the requirement gets people started working. But due to an increase in dropout, the overall effort with the requirement was no different from that without the requirement.

The paper is structured as follows. After reviewing the related literature, we present the experimental design in Section 2. Section 3 presents the theoretical model and predictions. In Section 4, we empirically compare daily and weekly goals and effort under them. Section 5 presents findings from the two extensions to our baseline treatments. Section 6 discusses alternative mechanisms and possible extensions. Section 7 concludes. The online appendix contains proofs, robustness checks regarding the theoretical predictions and the empirical analysis, a description of control variables, further results, and the experimental instructions.

**Related literature.** The narrow bracketing literature goes back to Tversky and Kahneman (1981). Much of it considers simultaneous risky choices, where narrow bracketing is a choice error (e.g. Rabin and Weizsäcker, 2009). We contribute to the literature strand that considers narrow bracketing as a tool to overcome self-control problems (e.g. Shefrin and

Thaler, 1988; Fudenberg and Levine, 2006). Our theoretical contribution is to provide the first model of how the bracketing of goals in repeated tasks affects the level of goals that people set for themselves, the actual effort provided, and the pattern of effort over time. Previous work gives conditions under which narrow bracketing is optimal with simultaneous tasks (Koch and Nafziger, 2016) or in a twice repeated optimal stopping problem (Hsiaw, 2018).<sup>3</sup> The phenomenon of effort substitution was previously noted in a simple two-period model by Jain (2009) and for two tasks by Koch and Nafziger (2016). These studies provide many important insights, but they do not capture situations where a task has to be performed repeatedly over some time and where the decision is not about stopping, but about how much effort to provide.

Our theoretical predictions build on the premise that people have a present bias in the real-effort task. Augenblick et al. (2015) are the first to estimate present bias in effort. They find evidence of present biased preferences in the effort domain, but not in the money domain. Further, they find that the present bias relates to demand for commitment devices. Augenblick and Rabin (2018) enrich the former study by eliciting the beliefs of the individuals about their future effort. They find that most subjects are (partially) naïve about their present bias.

Our empirical contribution is to compare behavior for narrowly and broadly bracketed goals in repeated tasks. Two early studies from psychology (Bandura and Simon, 1977; Bandura and Schunk, 1981) suggest that narrow ('proximal') goals are better at motivating effort than broad ('distal') goals. Yet, they have several conceptual problems and small sample sizes.<sup>4</sup> Other studies of goal setting in repeated tasks are distinct from our study because

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<sup>3</sup>Fischer and Ghatak (2016) study the effects of frequent vs. infrequent repayment instalments of loans in microfinance. Frequent repayment requirements have the disadvantage of delaying the reward (e.g. a new loan), but the advantage of providing better incentives because parts of the repayment are shifted to a future period. Neither effect is present with a narrow goal.

<sup>4</sup>In a 4-week weight loss program, Bandura and Simon (1977) assigned 27 subjects either to a condition with 'distal' goals for food consumption over one week or 'proximal' goals for each of four time periods during each day. Yet, more than half of the subjects apparently set proximal goals in the 'distal' treatment. While those subjects with proximal goals lost the most weight, the results must be interpreted with caution because they do not rely on exogenous variation. Bandura and Schunk (1981) ran a remedial program with children who had severe deficits in math. They assigned 10 subjects the narrow goal of completing a certain

they either focus on (repeated) daily goals only or have a single goal for the entire time span. In Kaur et al. (2015), workers could set individual, daily work targets that were then externally enforced by the firm with a penalty for low output. Kaur et al. observed the effort and daily goals of the workers repeatedly over a period of 13 months, but did not vary the goal bracket. Setting goals was a dominated option for rational workers because goals penalized low output without any additional reward for high output. Nevertheless, in 36 percent of the cases workers set positive goals. They produced more and earned more than when not offered the opportunity to set goals. Uetake and Yang (2018) provide descriptive evidence from a weight loss app that daily goals can help achieve a long-run goal. The app suggests daily calorie targets for achieving a self-chosen, long-run weight loss target. Other studies consider the impact of a single, self-set goal on the outcome of a repeated task, such as weight loss (Toussaert, 2016), energy saving (Harding and Hsiaw, 2014), or studying (van Lent and Souverijn, 2017; van Lent, 2018; Clark et al., 2019; Himmler et al., 2019). Most of these studies find a positive effect of goals.

By comparing goals and effort under different goal frames, we also contribute to the literature in economics that studies how to optimally design and set goals. Suvorov and van de Ven (2008), Jain (2009), Koch and Nafziger (2011), and Hsiaw (2013) model non-binding personal goals in one-time tasks. Goals help to overcome self-control problems by serving as reference points that make substandard performance psychologically painful. Empirical work yields mixed results on the impact of asking subjects to set goals in one-time tasks. While Akina and Karagozoglub (2017) observe that goals have no effect on performance, Smithers (2015) and Goerg and Kube (2012) find that goal setting increases performance. Other studies consider non-binding goals in work environments. These goals can either be tied to monetary rewards (Dalton et al., 2015; Goerg and Kube, 2012; Kaur et al., 2015) or not (Brookins et al., 2017; Corgnet et al., 2015, 2018). The latter find that goals increase performance. Evidence on the former is mixed. While Dalton et al. (2015) find an overall null result,

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number of problems in each of seven daily, 30-minute sessions. 10 other subjects were assigned the broad goal of completing the equivalent total number of problems over all sessions. Those assigned the fixed, narrow goals performed better in a number of dimensions than those assigned the fixed, broad goal. This study, however, has been criticized because only children in the narrow goal condition had been able to evaluate their progress toward their goal (cf. Kirschenbaum, 1985, p.494).

Goerg and Kube (2012) and Kaur et al. (2015) find positive effects. Overall, most studies point to goals having a positive impact on performance – specifically in the presence of monetary incentives. Several studies point to goals being more effective for men than for women (Smithers, 2015; Dalton et al., 2015; Clark et al., 2019) – a finding that we replicate (cf. Appendix J). Clark et al. (2019) find that effort goals outperform performance goals. Finally, we contribute to the literature on externally enforced commitment devices (for an overview see Bryan et al., 2010). Ariely and Wertenbroch (2002) observe that imposing binding deadlines on students improves their academic performance. Bisin and Hyndman (2014) and Burger et al. (2011), however, find no such effects. The novel feature of our experiment is to study whether a flexible, externally enforced minimum work requirement can complement self-enforced goals.

## 2 Experimental design and procedures

Our study included seven treatments, summarized in Table 1. In total 468 subjects participated (see Appendix E for details on the recruitment of subjects). We focus on the main treatments, *Daily* and *Weekly*, in Sections 2-4. In Section 5, we discuss four additional treatments that address certain mechanisms. One treatment (*NoManipulation*) does not contribute directly to the research question and is discussed in Appendix I. Experimental instructions are reproduced in Appendix M.

### 2.1 Design and treatments

Treatments consisted of a goal setting part and a work part. All parts were conducted online.

**Goal setting part.** On a Wednesday at midnight, subjects received an email informing them that they could earn up to 500 Danish kroner (\$83) by completing a short online questionnaire before Friday midnight and performing some online tasks in the following week from Monday to Friday. A reminder was sent out on Friday 9 am to those who had not responded.

Subjects completed a task based on Abeler et al. (2011). The task required them to count correctly the number of zeros in a series of tables with zeros and ones as in the following



Table 1: Overview of treatments.

	Treatment <sup>a</sup>	Goals set	Goal feedback	Min. work requirement <sup>c</sup>	N
<b>Sections 2-4</b>	Daily	daily	daily	no	78
	Weekly	weekly	weekly	no	77
<b>Section 5.1</b>	Daily(R) <sup>b</sup>	daily	daily	no	75
	Aggregated	daily	weekly	no	75
<b>Section 5.2</b>	DailyRequirement	daily	daily	yes	47
	WeeklyRequirement	weekly	weekly	yes	45
<b>Appendix I</b>	NoManipulation	no goals elicited		no	71

Notes. Total number of subjects 468. <sup>a</sup>Subjects got daily reminders about goals by email (except *NoManipulation*). <sup>b</sup>*Daily(R)* replicates *Daily*. <sup>c</sup>Complete at least one table per day.

one:

0	0	0	1	1
0	0	0	0	0
0	0	1	0	1
0	0	0	0	0
1	0	0	1	0
0	1	1	0	1

Figure 1: Example of a table for the counting task

Subjects had three minutes to complete as many tables as possible and earned DKK 0.5 (\$0.08) per completed table. If they miscounted the number of zeros in a table, an error message appeared. A table was not recorded as completed until the correct number was entered.<sup>5</sup> There was no punishment for miscounting. This stage ensured that subjects had a good understanding of how difficult the task was and provided us with a baseline measure of how easy the task was for a subject initially, referred to as *baseline productivity* in the following.

Subjects were then informed that they could complete up to 1,000 such tables at any time during the following week from Monday to Friday, and that they would receive DKK 0.5

<sup>5</sup>Subjects were not told that we allowed an error margin of  $\pm 1$ .

per completed table. Each day they would receive an email with a web link to complete the tables. They were asked a set of questions designed to make them think about the benefits of working on the task and their work week ahead.

Subjects completed the part by setting non-binding goals. The goal bracket differed across treatments. In *Daily*, subjects set for each day of the following week a separate goal for how many tables to complete (adding up to at most 1000 tables for the entire week). In *Weekly*, they set a weekly goal of up to 1000 for the number of tables to complete from Monday to Friday. Subjects were informed that we would remind them of their goals in the following week. On the final screen, subjects were told that they would receive an email at 0:00h on Monday with a link to the work screen.

**Work part.** In the following week, each day at 0:00h, subjects received an email with a link to the work screen. Subjects were reminded that they could work anytime until Friday 23:59h. In *Daily* (*Weekly*), the email additionally informed subjects about the goal they had set for that day (the week). The only other treatment difference was how the the first two lines above the table to be counted were presented (each table is on a separate screen). In *Daily* (*Weekly*), they showed the goal for the current day (week) and how many tables the subject had already counted on that day (during the week so far). Subjects always saw how many of the 1,000 tables remained, a reminder about the earnings, and a reminder that they could use the link to come back as often as they liked. Each time a subject completed a table, a new table appeared and the screen information was updated. If someone miscounted, an error message appeared and the same table was presented again. Upon reaching the maximum of 1,000 tables, a ‘Thank you’ screen appeared and no further counting was possible.

Our design aimed to create a work-leisure self-control problem. It featured generous pay to make it desirable to complete the task. Specifically, our pay was above the usual hourly wage for students of around DKK 130 per hour (completing all 1,000 tables required about 3 hours of work for DKK 500). But once a subject faced the task, its tedious nature made the leisure alternative tempting.

## 2.2 Procedures

Subjects were informed that payments would be made 2-6 weeks after the experiment by bank transfer via the Danish payment system through which public bodies and companies can send money to a person using their social security number. The procedure was required by Aarhus University and is perceived as normal by Danish citizens. The experiment ran online using the Qualtrics software.<sup>6</sup> It could be accessed via desktop, notebook, or touchpad. A software filter blocked access via smartphone. This was to avoid that subjects would solve a bit of the task here and there on their smartphone (say, while waiting for the bus), which might not be perceived as costly. Tables were copy protected to prevent pasting them into a spreadsheet program to do the counting. Sample sizes were determined by a rule of thumb, subject availability, and budget (Section 6 discusses power).

## 3 Theoretical predictions

Our theoretical predictions are based on a setting where a quasi-hyperbolic discounter (Laibson, 1997) works repeatedly on a task and can set daily or weekly goals to motivate himself. We first describe the model before we proceed to the theoretical predictions.

### 3.1 Model

**Task.** The individual works repeatedly on a task at  $t \in \{1, \dots, T\}$ . The activity requires effort  $e_t \in [0, \infty)$ , causing immediate costs  $c(e_t)$  (strictly increasing and strictly convex) and long-run benefits  $b(e_t)$  (strictly increasing and concave).

**Preferences.** We assume that the individual is a quasi-hyperbolic discounter. Self  $t$  (the incarnation of the individual at date  $t \in \{0, 1, \dots, T+1\}$ ) has utility  $U_t = u_t + \beta \left[ \sum_{\tau=t+1}^{T+1} u_\tau \right]$ , where  $u_t$  is the instantaneous utility. In the absence of goal setting, the instantaneous utility is given by  $u_0 = 0$ ,  $u_t = -c(e_t)$  for  $t \in \{1, \dots, T\}$ , and  $u_{T+1} = \sum_{t=1}^T b(e_t)$ . The present bias parameter  $\beta \in (0, 1)$  captures the extent to which the individual overemphasizes immediate

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<sup>6</sup>Patterns in IP addresses suggest that task outsourcing (e.g. to MTurkers) did not occur (Appendix G).

utility flows relative to future utility flows. The exponential discount factor  $\delta$  is set to one.<sup>7</sup> The present bias causes a work-leisure self-control problem. Self 0 weighs equally future costs and benefits and thus prefers effort to equate marginal costs and benefits for all dates:

$$b'(e_0^*) = c'(e_0^*). \quad (1)$$

Each self  $t \geq 1$  discounts future benefits by  $\beta < 1$ . So for all  $t = 1, \dots, T$ , self  $t$  prefers effort such that

$$\beta b'(e_t^*) = c'(e_t^*). \quad (2)$$

As  $\beta < 1$ , self 0 wants a higher effort than self  $t$ :  $e_0^* > e_t^*$ . To overcome this self-control problem, self 0 sets effort goals: Either in the form of a narrow goal  $g_t$  for each day  $t \in \{1, \dots, T\}$ , or a broad goal  $G$  for the sum of effort over all  $T$  days. In the context of our experiment,  $T = 5$  and treatment *Daily* elicits daily goals, whereas treatment *Weekly* elicits a weekly goal.

We allow the individual to hold an overly optimistic belief about his present bias  $\hat{\beta} \geq \beta$  (O'Donoghue and Rabin, 1999), encompassing the cases of sophistication ( $\hat{\beta} = \beta$ ), partial naïveté ( $\hat{\beta} \in (\beta, 1)$ ), and full naïveté ( $\hat{\beta} = 1$ ).

**Goals as reference points.** Consistent with the evidence from psychology on goals (e.g. Heath et al., 1999; Locke and Latham, 2002; Wu et al., 2008), we assume that future selves take their goals as reference points (for a model that allows for goal revision see Koch and Nafziger, 2016). With narrow goals, the individual compares the actual effort  $e_t$  with the goal  $g_t$ . With a broad goal, the individual compares the overall effort  $\sum_{t=1}^T e_t$  with the goal  $G$ . If effort differs from the goal by  $z$ , the individual experiences a corresponding comparison utility  $\mu(z) = z$  for  $z < 0$ , and  $\mu(z) = 0$  for  $z \geq 0$ .<sup>8</sup> The individual experiences the comparison utility in the last period when the benefits accrue (this assumption can be relaxed; see Appendix D). That is, with a broad goal, we have  $u_{T+1} = \sum_{t=1}^T b(e_t) + \min\{0, \sum_{t=1}^T e_t - G\}$ , and with narrow goals we have  $u_{T+1} = \sum_{t=1}^T [b(e_t) + \min\{0, e_t - g_t\}]$ .

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<sup>7</sup>Using a real-effort task similar to ours, Augenblick and Rabin (2018) find that subjects are present-biased and have an estimated daily discounting parameter  $\delta \approx 1$ .

<sup>8</sup>Defining comparison utility over effort matches the frame of the experiment. For an alternative approach that assumes that the goal induces reference standards for costs and benefits, see Koch and Nafziger (2016).

**Equilibrium.** The equilibrium concept is that of preferred personal equilibrium (Kőszegi and Rabin, 2006). Goals are assumed to be rational in the sense of perception perfection (O'Donoghue and Rabin 1999; 2001). The goal(s) that self 0 sets have to be consistent with the (possibly erroneous) beliefs  $\hat{e}_{t,0}$  that self 0 holds about his future effort at dates  $t = 1, \dots, T$ . That is,  $\hat{e}_{t,0} = g_t$  for narrow goals and  $\sum_{t=1}^T \hat{e}_{t,0} = G$  for a broad goal.

## 3.2 Analysis: Sophisticated individual

We characterize the implementable effort profiles and optimal goals for daily and weekly goal setting formats. In the next section, we discuss what changes when we introduce naïvité.

### 3.2.1 Daily goals (narrow goals)

**Implementable effort profiles.** To characterize the effort levels that self 0 can implement with daily goals, we need to ask when his future self, who works on task  $t$ , does not have an incentive to deviate from goal  $g_t$ . If self  $t$  puts in at least the effort required by his goal, his utility is  $\beta b(e_t) - c(e_t)$ . If effort falls short of the goal, he suffers a loss and the utility after a deviation  $e_t < g_t$  is:

$$\beta b(e_t) - c(e_t) - \beta (g_t - e_t). \quad (3)$$

For a goal to be implementable, the utility from sticking to the goal has to exceed the utility from falling short of it. That is, (3) has to be increasing in  $e_t$  for any  $e_t < g_t$ . This is the case for any goal that is not ‘too high’, i.e., that does not exceed  $e_{max}(\beta)$  defined by

$$\beta [b'(e_{max}(\beta)) + 1] = c'(e_{max}(\beta)). \quad (4)$$

The maximal implementable effort  $e_{max}(\beta)$ , defined by (4), is increasing in  $\beta$ . Further, note that (2) and (4) imply  $e_{max}(\beta) > e_t^*$ . The maximal implementable effort exceeds the preferred goal of self  $t$  because the fear of a loss makes self  $t$  strive harder than he would in the absence of comparison utility. Similarly, the effort cannot fall short of  $e_t^*$ , because self  $t$  will always choose at least this effort level.

**Goal setting.** Self 0 picks his daily goals to maximize his utility subject to the goal being implementable. The following result summarizes the findings. Proofs are in Appendix A.

**Proposition 1** Suppose a sophisticated individual ( $\hat{\beta} = \beta$ ) sets daily goals. Self 0 picks the implementable goal that maximizes his utility:  $\max_{g_t \in [e_t^*, e_{max}(\beta)]} \beta [b(g_t) - c(g_t)]$ .

1. For  $\beta$  large enough,  $e_{max}(\beta) \geq e_0^*$ . Self 0 sets daily goals equal to his preferred effort,  $g_t^* = e_0^*$  for  $t = 1, \dots, T$ , and self  $t$  provides  $e_t = g_t^*$ .
2. For lower values of  $\beta$ ,  $e_{max}(\beta) < e_0^*$ . Self 0 sets daily goals  $g_t^* = \min\{e_0^*, e_{max}(\beta)\}$  and selves  $t = 1, \dots, T$  provide effort  $e_t = g_t^* = e_{max}(\beta)$ , where  $e_0^* > e_t > e_t^*$ . That is, effort exceeds the effort that self  $t$  would pick in the absence of comparison utility but still lies below the level  $e_0^*$  that self 0 would prefer.

### 3.2.2 Weekly goal (broad goal)

**Implementable effort profiles.** A first insight is that a weekly goal cannot improve self-regulation relative to daily goals. This is because the incentives to deviate from the goal in a single period are the same under a weekly goal as under daily goals – provided all other selves stick to the plan. Thus,  $e_{max}(\beta)$  defined in (4) is also the maximal implementable effort in a given period with a weekly goal. If self  $t$  believes that the weekly goal can still be reached with some  $e_t \leq e_{max}(\beta)$ , he provides at least such an effort. If self  $t$  believes that  $G$  no longer will be achieved for any  $e_t \leq e_{max}(\beta)$ , he provides  $e_{max}(\beta)$  (see Lemma 1 and its proof in Appendix A.2).

But a weekly goal can harm self-regulation. Self 0 may not be able to implement certain effort profiles with a weekly goal that are implementable with a daily goal, because future selves would deviate, for example, by lowering effort today and compensating with increased effort tomorrow. We refer to such behavior as *effort substitution*. To provide some intuition, suppose self 0 sets a weekly goal that equals the sum of his desired daily effort,  $G = T e_0^*$ , and  $e_0^* < e_{max}(\beta)$ . Further, suppose that all selves provided  $e_0^*$ , except that self  $T - 1$  worked less hard than  $e_0^*$ . Now if self  $T$  just provided  $e_0^*$ , the individual would suffer a loss from falling short of  $G$ . To avoid this loss, self  $T$  will increase his effort up to  $e_{max}(\beta)$ . For self  $T - 1$  it indeed pays off to work less than  $e_0^*$ , knowing that self  $T$  will work harder to make up for the shortfall. Because of the present bias, self  $T - 1$  prefers (on the margin) to shift effort costs into the future. Consequently, the individual does not provide  $e_0^*$  in every period when facing the broad goal  $G = T e_0^*$  (see also Proposition 4 in Appendix A.3.3). That is, even though

$e_0^* < e_{max}(\beta)$  self 0 cannot implement his preferred daily effort  $e_0^*$  in each period. Instead, he has to implement an increasing effort profile that prevents him from pushing (even more) effort to the future. In contrast, if  $e_0^* = e_{max}(\beta)$ , each self  $t \in \{1, \dots, T-1\}$  is committed not to lower his effort because future selves will not compensate. The next result summarizes and extends these insights for any weekly goal  $G \leq T e_{max}(\beta)$ . The effort profile can either be increasing,  $e_1 \leq e_0^* < e_T$  (with  $e_T < e_{max}(\beta)$  or a corner solution  $e_T = e_{max}(\beta)$ ), or flat with  $e_t = e_{max}(\beta)$  in all periods  $t = 1, \dots, T$ .

**Proposition 2** *Suppose a sophisticated individual ( $\hat{\beta} = \beta$ ) sets a weekly goal  $G \leq T e_{max}(\beta)$ . The goal is achieved:  $G = \sum_{t=1}^T e_t$ .*

1. *For  $e_{max}(\beta) > e_0^*$ , effort is (weakly) increasing over time:  $e_1 \leq e_0^* < e_T$ . There exists  $\underline{t} \in \{2, \dots, T+1\}$  such that  $e_1 < e_2 < \dots < e_{\underline{t}-1} < e_{\underline{t}} = e_{\underline{t}+1} = \dots = e_T = e_{max}(\beta)$ . If  $\underline{t} > 2$ ,  $e_1 < e_0^* < e_{\underline{t}-1}$ . If  $\underline{t} = 2$ ,  $e_1 = e_0^*$ .*
2. *For  $e_{max}(\beta) \leq e_0^*$ ,  $G^* = T e_{max}(\beta)$ ,  $e_t = e_{max}(\beta)$  in all periods  $t = 1, \dots, T$ .*

**Goal setting.** How does effort substitution affect the goal that self 0 sets? Our next result imposes a technical assumption on the third derivatives of the benefit and cost functions, which is satisfied, for example, if the benefits are linear and costs are quadratic.<sup>9</sup> Under this assumption, we can compare the optimal weekly goal to the optimal daily goals. For  $e_{max}(\beta) \leq e_0^*$ , there is no difference between the two goal setting formats and  $G^* = T e_{max}(\beta)$  (Proposition 2). So we focus on the case where  $e_{max}(\beta) > e_0^*$ .

**Proposition 3** *Suppose a sophisticated individual ( $\hat{\beta} = \beta$ ) sets a weekly goal,  $\beta b'''(e) - c'''(e) \geq 0$  and  $b'''(e) - c'''(e) \geq 0$ . Define  $\check{\beta} : e_{max}(\check{\beta}) = e_0^*$ .*

1. *There exists  $\bar{\beta} \in (\check{\beta}, 1)$ , such that for  $\beta \geq \bar{\beta}$  self 0 sets a total goal  $G^* < T e_0^*$  and anticipates interior effort  $\hat{e}_{t,0} < e_{max}(\beta)$  for every period  $t = 1, \dots, T$ .*
2. *There exists  $\underline{\beta} \in (\check{\beta}, \bar{\beta})$ , such that for  $\beta \in [\underline{\beta}, \bar{\beta}]$  self 0 sets  $G^* = (T-1) e_{max}(\beta) + e_0^*$  and anticipates a corner solution for periods  $t > 1 : (e_0^*, e_{max}(\beta), \dots, e_{max}(\beta))$ .*

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<sup>9</sup>Augenblick and Rabin (2018) find an approximately quadratic cost function when estimating effort costs for a real-effort task similar to ours.

Given that  $e_{max}(\beta) > e_0^*$ , self 0 could in principle implement the sum of desired daily effort levels with a weekly goal  $G = T e_0^*$  and thereby achieve the same overall effort as with daily goals  $g_t = e_0^*$ . Yet, because of effort substitution, effort would be asymmetrically allocated over the days, starting below and ending above the desired daily effort of self 0. Because of the strictly convex effort cost, the utility of self 0 under the weekly goal  $G = T e_0^*$  is lower than the utility under the equivalent daily goals  $g_t = e_0^*$ . Part 1 of Proposition 3 shows that self 0 chooses a lower weekly goal than  $T e_0^*$  if  $\beta$  is sufficiently large. The reason is that lowering the goal relative to  $T e_0^*$  reduces the spread in effort costs across periods and leads to a lower average cost per unit of effort. Part 2 of Proposition 3 shows that for a relatively severe present bias (low  $\beta$ ) it may pay to commit at least self 1 to provide  $e_0^*$  by setting a weekly goal that forces all selves  $t > 1$  to provide  $e_t = e_{max}(\beta)$ . Essentially, this follows from the continuity of  $e_{max}(\beta)$ : If  $\beta$  is such that  $e_{max}(\beta)$  only slightly exceeds  $e_0^*$ , the cost of excessive effort  $e_{max}(\beta)$  is negligible and the utility is close to the self 0 optimum. Note that part 1 corresponds to the case where  $\underline{t} = T + 1$  in Proposition 2 (self 0 anticipates interior effort  $\hat{e}_{t,0} < e_{max}(\beta)$  for every period  $t = 1, \dots, T$ ) and part 2 corresponds to the case where  $\underline{t} = 2$ . These two cases are the essential ones for understanding the main driving forces. The other possibilities of partial corner solutions  $\underline{t} \in \{3, \dots, T\}$  are included in the proof and show up as a ‘staircase’ pattern in the left panel of Figure 2 for our parametric example.

### 3.3 Analysis: Partially naïve individual

We next discuss the implications of partial naïveté. We make two assumptions. First, the individual cannot revise the goal. Second, given past effort and his wrong belief about the present bias, the individual is able to update beliefs about future effort and goal achievement. We discuss these assumptions in Section 6.

#### 3.3.1 Daily goals

A partially naïve self 0 picks his daily goals for tasks to maximize his utility  $\hat{\beta} [b(g_t) - c(g_t)]$  subject to  $g_t \in [e_t^*, e_{max}(\hat{\beta})]$ . The individual overestimates what goals are realistic, because the perceived maximal implementable effort  $e_{max}(\hat{\beta})$  exceeds the actual  $e_{max}(\beta)$  for  $\hat{\beta} > \beta$  (for details see Appendix A.1). Yet, this mistake has no consequences as long as the goal



satisfies  $g_t \leq e_{max}(\beta)$ . In this case, the goal is achieved. If the individual sets an overly ambitious goal  $g_t > e_{max}(\beta)$ , self  $t$  will underperform relative to that goal.

### 3.3.2 Weekly goal

A partially naïve individual sets weekly goals in the same way as a sophisticated individual, except that the individual applies  $\hat{\beta}$  (the belief about his present bias) when predicting future behavior rather than  $\beta$  (his true present bias). As a consequence, in Proposition 3, whether case 1 or case 2 applies does not depend on the actual  $\beta$ , but on the belief  $\hat{\beta}$ . For example, self 0 chooses a weekly goal  $G < T e_0^*$  if  $\hat{\beta}$  is sufficiently large.

However, the actual effort pattern differs from the one that self 0 anticipates. First, self 0 anticipates that  $\hat{\beta}$  will be applied whereas, when deciding on the effort  $e_\tau$ , self  $\tau > 0$  applies the actual  $\beta < \hat{\beta}$ . Second, while all selves hold the same wrong belief that future selves will provide effort up to  $e_{max}(\hat{\beta})$ , at some point the individual will observe a different history of past effort than self 0 anticipated. At that point, self  $\tau$  will plan to compensate for the short-fall and beliefs will have to shift upward to adjust for lagging behind the original expectations. As a result, actual effort falls short of the one-period ahead expectation and expectations about future effort increase relative to those held in the previous period.<sup>10</sup> For a formal statement and proof of this result, see Proposition 5 in Appendix A.6.

Naïvité exacerbates the problem of effort substitution. Because the individual incorrectly predicts the extent to which a future self will increase his effort in response to him providing less effort today, the individual might lower his effort ‘too much’ and fail to meet the goal. Note that the individual believes that his future selves will provide effort up to  $e_{max}(\hat{\beta})$ . So even if the goal is no longer achievable, the individual will not realize this until the point comes when the goal cannot be achieved even with  $e_{max}(\hat{\beta})$  in every following period. In our parametric example in Figure 2, for example, the individual only realizes in the final period that he will fail the goal.

Naïvité also has consequences for goal achievement. A partially naïve individual is weakly more likely to fail the weekly goal compared to a situation where he sets daily goals. For  $e_{max}(\beta) < e_0^*$ , the individual fails to reach his goals under both goal setting formats. For

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<sup>10</sup>Except in the case of a corner solution where self 0 expects effort  $\hat{e}_{1,0} = e_0^*$  for period 1 and  $\hat{e}_{t,0} = e_{max}(\hat{\beta})$  for periods  $t = 2, \dots, T$ . Here actual effort in period 1 may match expectations:  $e_1 = \min\{e_0^*, e_{max}(\beta)\}$ .

$e_{max}(\beta) \geq e_0^*$ , daily goals would be achieved ( $g_t^* = e_0^* = e_t$ ). Here the weekly goal may not be achieved though: The individual only achieves the goal if self  $T$  faces a remaining part of the goal  $G_T = G - \sum_{t=1}^{T-1} e_t \leq e_{max}(\beta)$ . Clearly, this is violated in case of a corner solution where self 0 anticipates effort in period  $T$  of  $\hat{e}_{T,0} = e_{max}(\hat{\beta})$ , because  $e_{max}(\hat{\beta}) > e_{max}(\beta)$ . In case of an interior solution with  $\hat{e}_{T,0} < e_{max}(\hat{\beta})$ , we know from Proposition 3 that self 0 anticipates  $\hat{e}_{T,0} > e_0^*$ . In addition, because of naïveté, the actual effort falls short of the one-period ahead expected effort. Consequently, expectations about future effort increase. This further pushes-up the remaining goal  $G_T$ . So even though self 0 sets what looks like a ‘feasible’ goal ( $G^* < T e_0^* < T e_{max}(\beta)$ ), self  $T$  faces  $G_T > e_0^*$  which may exceed  $e_{max}(\beta)$ .

### 3.4 Parametric example

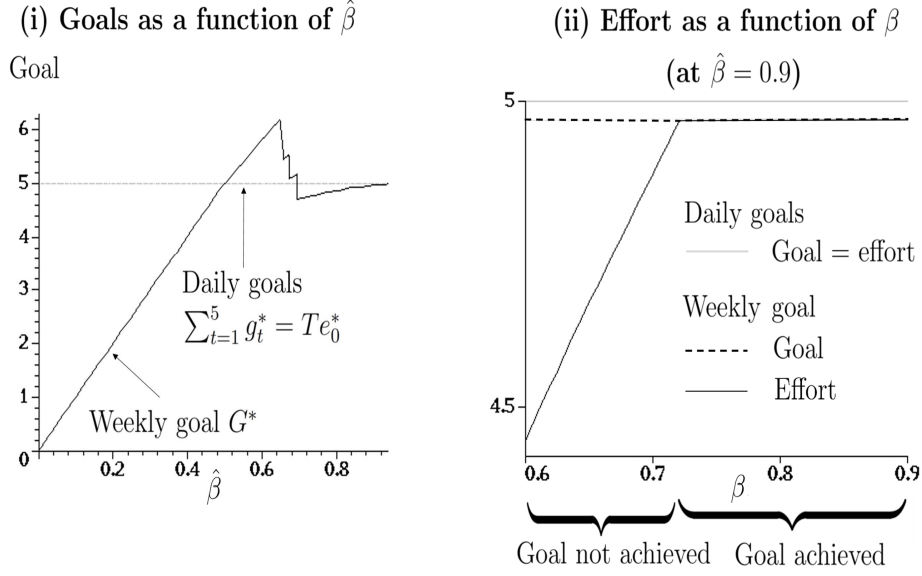
To illustrate which cases are most relevant we provide a parametric example with  $b(e) = e$  and  $c(e) = \frac{1}{2} e^2$ . Appendix B provides the detailed calculations. We have  $e_0^* = 1$ ,  $e_1^* = \beta$ , and  $e_{max}(\beta) = 2\beta$ . With daily goals, self 0 sets as goal his preferred effort as long as  $\hat{\beta} \geq \frac{1}{2}$ , because then  $e_0^* \leq e_{max}(\hat{\beta})$ . The left panel of Figure 2 plots the optimal weekly goal  $G^*$  as a function of  $\hat{\beta}$  if  $T = 5$ , like in our experiment. For  $\hat{\beta} \leq \frac{1}{2}$  the individual implements  $e_{max}(\hat{\beta})$ , as he would for daily goals. For  $\hat{\beta} \in (\frac{1}{2}, \bar{\beta})$ , where  $\bar{\beta} \approx 0.694$ , the individual has a higher weekly goal than the sum of daily goals would have been, because of a corner solution like in Part 2 of Proposition 3. For  $\hat{\beta} \geq \bar{\beta}$ , Part 1 applies and the individual has a lower weekly goal than the sum of daily goals would have been.

Augenblick et al. (2015) estimate a population present bias parameter of  $\beta = 0.9$  using a real-effort task similar to ours and find that almost all mass of present-biased individuals lies on  $\beta \in [0.6, 1)$  (see their Figure VI).<sup>11</sup> Augenblick and Rabin (2018) estimate  $\beta = 0.83$  and

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<sup>11</sup>Some of our subjects participated in a prior study (Epper et al., 2018), where we implemented the elicitation task of Augenblick et al. (2015). We estimate a population  $\beta$  of 0.94, restricting our sample to those subjects whose choices were monotonic. Compared to Augenblick et al. (2015), a larger proportion of subjects in our data make choices that are non-monotonic in the ‘efficiency’ of effort (conversion rate between effort today vs. effort in one week). Around 45 percent of subjects are present biased, 15 percent dynamically consistent, and 40 percent future biased, compared to 33, 47, and 21 percent, respectively in Augenblick et al. (2015). The small number of observations where we are able to estimate an individual  $\beta < 1$  prevents us from exploiting this measure in our empirical analysis.

Figure 2: Goals and effort in the parametric example with  $T=5$ .



Notes.  $b(e) = e$  and  $c(e) = \frac{1}{2} e^2$ , where  $e_0^* = 1$ ,  $e_1^* = \beta$  and  $e_{max} = 2\beta$ .

$\hat{\beta} \approx 1$ . Individual estimates for present-biased individuals are concentrated on  $\beta \in [0.5, 1)$  and  $\hat{\beta}$  is concentrated around 1 (see their Figure 6). They find only a weak correlation between  $\beta$  and  $\hat{\beta}$ .

In the right panel of Figure 2, we therefore consider a partially naïve individual with  $\hat{\beta} = 0.9$  and the actual effort response for  $\beta \in [0.6, 1]$ . Note that the goals for a partially naïve individual depend on  $\hat{\beta}$  and do not vary with  $\beta$ . With daily goals, the individual would always achieve his goal and the implemented effort would equal the preferred effort of self 0 ( $\sum_t^T g_t^* = \sum_t^T e_t = T e_0^* = 5$ ). For a weekly goal, the goal  $G^*$ , and hence the effort, are always below what they would have been with daily goals. The weekly goal is achieved except in the case where the individual severely overestimates  $\beta$ . Non-achievement occurs for  $\beta < \tilde{\beta}$ , where  $\tilde{\beta} \approx 0.722$ .

To sum up, the evidence on  $\beta$  and  $\hat{\beta}$  from Augenblick et al. (2015) and Augenblick and Rabin (2018) indicates that most people have  $\hat{\beta}$  close to 1 and that few people have a severe present bias. This suggests that Part 1 of Proposition 3 is the relevant case for most subjects, i.e., subjects set a lower goal and provide less effort under a weekly goal than under daily goals. Further, few subjects are likely to so severely overestimate  $\beta$  that they drive a wedge

between *Weekly* and *Daily* in terms of goal achievement. With these observations, we reach the following hypotheses:

### **Hypothesis 1 (Goals and effort)**

**H1a** *The total goal in Daily is larger than in Weekly.*

**H1b** *The total effort in Daily is larger than in Weekly. Controlling for the total goal, the total effort in both treatments is equal.*

### **Hypothesis 2 (Goal achievement)**

**H2a** *The goal achievement rates in Daily and Weekly are equal.*

**H2b** *Goal non-achievers are closer to their total goal in Daily than in Weekly.*

### **Hypothesis 3 (Effort substitution)**

**H3a** *The effort on Monday in Daily is higher than in Weekly.*

**H3b** *The effort on Friday in Daily is lower than in Weekly.*

## **4 Empirical findings: Daily goals vs. weekly goal**

This section analyzes the observed treatment differences between *Daily* and *Weekly*, considering in turn goals, effort, goal achievement, and effort substitution. For each hypothesis, we present descriptive statistics, non-parametric tests, and use regression analysis to test for treatment differences. Tables and figures prefaced with ‘A’ are in the online appendix.

**Estimation procedure.** Our design aimed to make high effort desirable from self 0 perspective, which led to substantial censoring of goals and effort at the maximum of 1,000 tables.<sup>12</sup> We deal with censoring using tobit regressions. Robustness checks indicate that results are robust to relaxing assumptions of the tobit model (Appendix L).

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<sup>12</sup>Roughly half of the subjects in *Daily* and *Weekly* chose a goal of completing all 1,000 tables. 30 and 41 percent of subjects, respectively, completed all 1,000 tables (Table 2).

Table 2: Descriptive statistics.

Treatment	N	Average total		Fraction of subjects with			Average		
		goal	effort	goal =1000	effort =1000	effort <goal	effort -goal	number of logins <sup>a</sup>	effort per login
Daily	78	789 (304)	690 (370)	0.51	0.41	0.45	-98.54 (399.47)	6.72 (4.78)	131.45 (109.63)
Weekly	77	682 (336)	521 (392)	0.47	0.30	0.47	-161.01 (398.14)	5.52 (4.32)	104.06 (97.90)

Notes. Standard deviations in parentheses. <sup>a</sup>New login: new day or >30 min. since last entry.

We regress the respective outcome variables on *daily goals*, which is a dummy for the treatment being *Daily*. The following primary outcome variables are available for all 155 subjects in *Daily* and *Weekly*. To test Hypothesis H1a (Section 4.1), we use the *total goal* (the aggregated daily goals in *Daily*, the weekly goal in *Weekly*). To test H1b (Section 4.2), we use the *total effort* (the total number of correctly counted tables). To test H2a,b (Section 4.3), we use *goal achievement* (a dummy variable whether  $total\ effort \geq total\ goal$ ). Finally, to test for the effort substitution hypotheses H3a,b (Section 4.4), we use *effort* on a given weekday, with specific focus on Monday and Friday. Secondary outcome variables, such as logins and task completion time allow us to study some mechanisms (Sections 5 and 6).

Control variables available for all treatments are a *gender* dummy and *baseline productivity* (balance tests are in Table A4).<sup>13</sup> Data from a prior study (Epper et al., 2018) provide us with a wider range of control variables (*full set of controls*), explained in Appendix F. For some of the extension treatments, these are not available though.

<sup>13</sup>Before subjects set goals, we asked them how many hours they had available for the task in the following week, how many tables they thought they could realistically solve within this time, and how much money they wanted to earn. These questions were meant to make subjects think about the task and their goals. They are collinear with the goals that subjects set and therefore not used as control variables.

## 4.1 Goals

We test Hypothesis H1a by comparing the total goal in *Daily* and *Weekly*. In line with the hypothesis, subjects in *Daily* set a total goal of 789 tables on average, whereas those in *Weekly* set a goal of 682 (Table 2; Permutation test for two independent samples,  $p = 0.041$ ).<sup>14</sup>

In tobit regressions, this difference is significant only after controlling for gender and productivity – indicating that the coefficients are imprecisely estimated without the controls (Table 3). Looking at the distribution of goals shown in the left panel of Figure 3, *Daily* had more mass on high goals than *Weekly*. For example, 65 percent aimed for at least 800 tables in *Daily* compared to 48 percent in *Weekly*. This can also be seen in the left panel of Figure 4, which plots the quantiles of goals. All quantiles below the median appear to be shifted upward in *Daily* relative to *Weekly*. Quantile regressions (controlling for baseline productivity and gender; with bootstrapped standard errors, 1000 replications) reveal significant treatment effects for the 20th and 30th percentile of 172.18 ( $se = 88.51$ ,  $p = 0.054$ ) and 245.10 ( $se = 116.44$ ,  $p = 0.037$ ), respectively. For the 40th and 50th percentile, we obtain 161.11 ( $se = 121.24$ ,  $p = 0.186$ ) and 40.4 ( $se = 112.76$ ,  $p = 0.721$ ), respectively.

## 4.2 Effort

First, we provide some descriptive statistics on the effort measure, and then we test Hypothesis H1b by comparing the total effort in *Daily* and *Weekly*.

Subjects took on average 15 seconds to complete a table and 90 percent of subjects required 20 seconds or less on average per table.<sup>15</sup> More productive subjects were slightly faster (OLS coefficient  $-0.390$ ,  $se = 0.124$ ,  $p = 0.002$ ). A one standard deviation higher baseline productivity cut completion time by 1.6 seconds. Subjects made 38 mistakes in the range of  $\pm 1$  around the correct solution on average. For technical reasons, we could not record mistakes that required recounting of a table.

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<sup>14</sup>A non-parametric test of difference in distributions. Permutation tests perform well even with censoring (Neuhaus, 1993). We ran Monte Carlo simulations with 10,000 repetitions, using Kaiser’s (2007) implementation `permtest2`. Permutation tests and Fisher’s exact tests reported below are all against the two-sided null hypothesis of no difference.

<sup>15</sup>We truncate completion time at 120 seconds to take out effects of breaks/stopping to log on again later.

Table 3: Impact of goal setting format on total goal and total effort (*Daily* vs. *Weekly*).

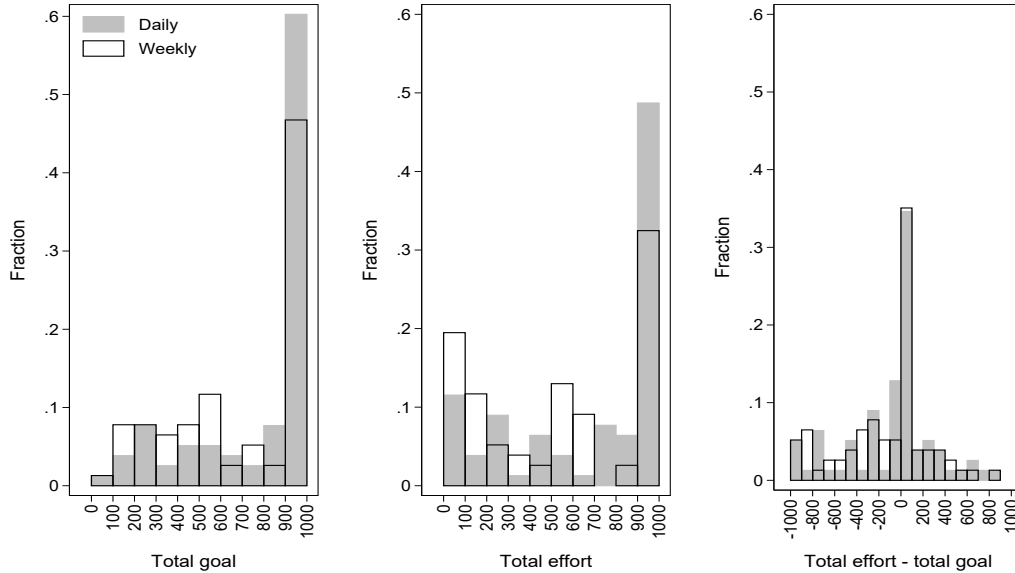
Dependent variable	Total goal				Total effort			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Daily goals <sup>a</sup>	150.95 (96.82)	173.14* (89.37)	186.89** (92.25)	235.46** (96.97)	228.39** (94.86)	180.91* (95.09)	151.44* (91.31)	100.89 (92.44)
Total goal							0.60*** (0.14)	0.60*** (0.15)
Baseline		24.17**	10.88		33.52***	67.04***	24.49**	57.82**
productivity		(11.08)	(19.00)		(10.67)	(23.59)	(10.38)	(22.41)
Female		-417.57*** (90.65)	-448.27*** (102.71)		-93.66 (99.86)	-176.73 (117.39)	53.56 (99.11)	-26.97 (114.60)
Constant	866.26*** (78.33)	677.16*** (184.81)	805.54 (562.15)	611.57*** (71.05)	109.27 (176.97)	-99.35 (645.24)	-230.05 (168.23)	-349.58 (616.96)
Full controls <sup>b</sup>	no	no	yes	no	no	yes	no	yes
Margin.effect(daily goals) <sup>c</sup>	75.14	89.44*	97.36**	135.79**	134.38**	110.07*	93.93*	64.15
Effect size <sup>d</sup>	0.22	0.27	0.29	0.35	0.34	0.33	0.24	0.19
N	155	155	153	155	155	153	155	153

Notes. Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis.

\* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>See Appendix F. <sup>c</sup>Tobit marginal effect on the censored latent variable

(at the means of control variables). <sup>d</sup>  $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly}}$ .

Figure 3: Distribution of goals and effort (*Daily* & *Weekly*).



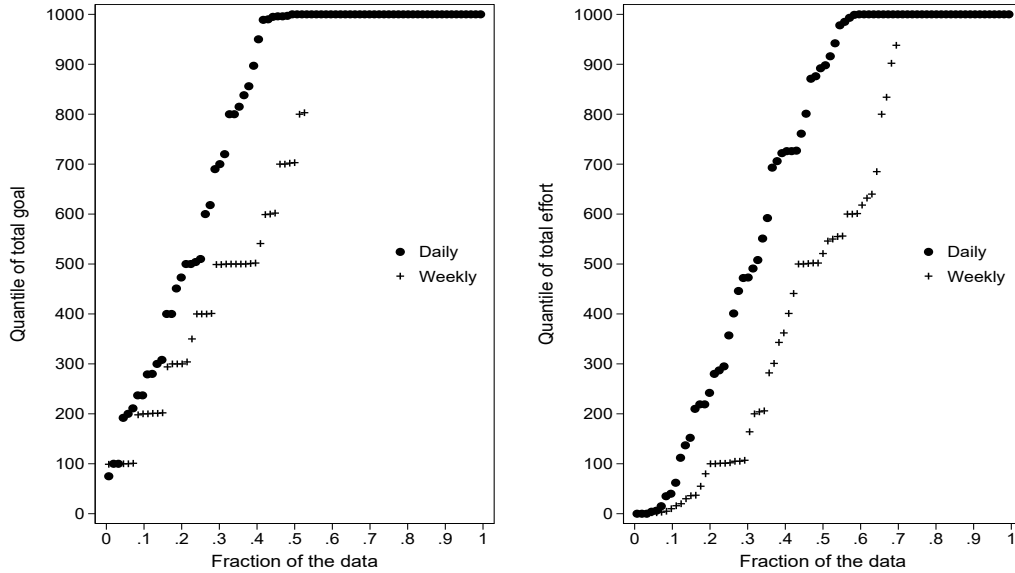
In line with Hypothesis H1b, subjects provided more effort in *Daily* than in *Weekly* on average: 690 vs. 521 tables (Table 2; permutation test,  $p = 0.005$ ). This difference is significant in tobit regressions (Table 3) and is also reflected in the distribution of effort. The middle panel of Figure 3 shows that *Daily* has more mass on high effort than *Weekly*. In the same vein, the right panel of Figure 4 shows that quantiles shift upward in *Daily* relative to *Weekly*. Quantile regressions reveal (borderline) significant treatment effects for the 20th, 30th, 40th, and 50th percentile of 209.33 ( $se = 101.08$ ,  $p = 0.040$ ), 223.00 ( $se = 35.18$ ,  $p = 0.101$ ), 230.71 ( $se = 134.06$ ,  $p = 0.087$ ), and 273.31 ( $se = 123.60$ ,  $p = 0.029$ ), respectively.

As the goal is endogenous, the effort gap between treatments either could be driven by the different goal frame or by differences in goal setting or goal achievement. We examine the latter two explanations next.

**Controlling for the total goal.** Once we control for the total goal, the treatment difference in effort becomes smaller and is no longer consistently significant (Table 3). The effort gap between *Daily* and *Weekly* appears mainly to be related to the higher total goal in *Daily* compared to *Weekly*, in line with Hypothesis H1b. Another indication that *Daily* increased effort by exogenously shifting goals upwards relative to *Weekly* is that the instrumental vari-



Figure 4: Quantiles of goals and effort (*Daily* & *Weekly*).



ables tobit coefficient in a 2SLS regression is larger than the regular tobit coefficient on total goal (Tobit 0.66,  $se = 0.14$ ,  $p < 0.001$ ; IV 1.89,  $se = 0.92$ ,  $p = 0.039$ ).

### 4.3 Goal achievement

In this section, we test for treatment differences in goal achievement (Hypotheses H2a and H2b). According to our model, naïve individuals overestimate the maximum amount of effort they will be able to put in. However, as summarized in Hypothesis H2a, this should typically not lead to a treatment difference in goal achievement. Indeed, we find no treatment difference in goal achievement (Fisher’s exact test,  $p = 0.872$ ; Table A5).

In addition, according to Hypothesis H2b, goal non-achievers are predicted to be closer to their total goal in *Daily* than in *Weekly*. Indeed, the right panel of Figure 3 shows that more mass is concentrated just to the left of zero in *Daily* than in *Weekly*. Tobit regressions of *total effort*-*total goal* for those subjects who fell short of the total goal (but attempted working) reveal a significant treatment effect once controlling for gender and productivity (Table A6).<sup>16</sup>

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<sup>16</sup>Because,  $e_{max}(\beta) > 0$  for  $\beta > 0$ , the argument behind Hypothesis H2b only applies if any attempt at effort is made ( $total\ effort > 0$ ). This excludes three subjects from each treatment, with an average goal of 999 for *Daily* and 734 for *Weekly*.

Table 4: Transition from goal profiles to effort profiles in *Daily*.

Goal profiles	Effort profiles							
	Flat <sup>d</sup>		High-low <sup>e</sup>		Low-high <sup>f</sup>		Other	
	N	Percent	N	Percent	N	Percent	N	Percent
Flat <sup>a</sup>	1	5.88	12	70.59	3	17.65	1	5.88
High-low <sup>b</sup>	0	0.00	24	77.42	5	16.13	2	6.45
Low-high <sup>c</sup>	0	0.00	13	48.15	13	48.15	1	3.70
Other	0	0.00	2	66.67	1	33.33	0	0.00

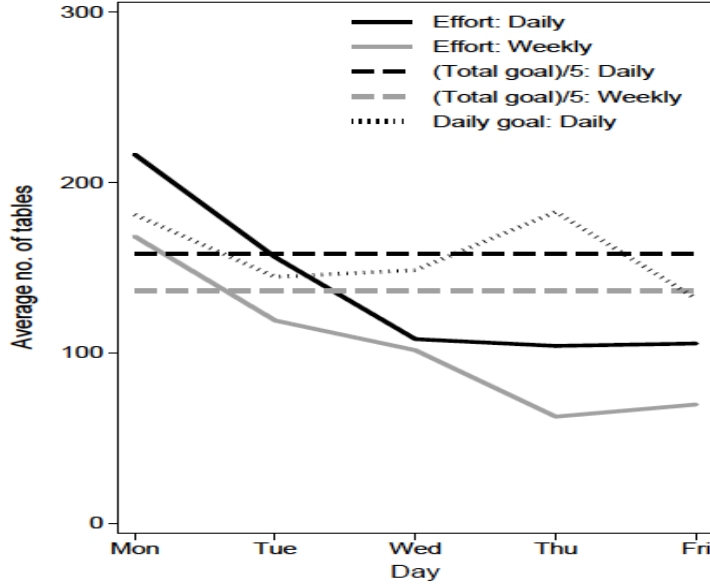
Notes. <sup>a</sup> $g_{mon} = g_{tue} = \dots = g_{fri}$ . <sup>b,c</sup> $g_{mon} + g_{tue} >^b (<^c) g_{thu} + g_{fri}$ . <sup>d,e,f</sup> Analogous to goal profiles.

#### 4.4 Effort substitution

We examine the emergence of effort substitution by looking at effort patterns over time. Specifically, we test Hypothesis H3a (H3b) by comparing effort on Monday (Friday) for *Daily* vs. *Weekly*. Figure 5 reveals that the average daily goals in *Daily* exhibited a downward sloping pattern (also found in the other treatments with daily goals discussed in Section 5). Subjects might have wanted to work less at the end of the week, because student parties or other leisure options increased opportunity costs. Extending our model with marginal costs that increase in  $t$  leads to a downward sloping pattern of goals for *Daily*, because both the self-0 preferred effort for date  $t$ ,  $e_{t,0}^*$ , and the maximal implementable effort,  $e_{max,t}(\beta)$ , then are decreasing in  $t$ . For *Weekly*, this extension affects effort substitution (Hypotheses H3a,b). While effort on Monday still is likely to be lower in *Weekly* than in *Daily*, the prediction of higher effort on Friday in *Weekly* than in *Daily* is weakened because a corner solution with  $e_T = e_{max,T}(\beta)$  in both treatments is more likely (see Appendix C).

The data from *Daily* reveal some heterogeneity in goal profiles (Table A7). 86 percent of subjects set a non-zero goal for each day, 22 percent set the same goal for each day, 40 percent aimed to ‘start high and end low’ (the average goal for the first two days of the week exceeded the average goal for the last two days), and 35 percent aimed to ‘start low and end high’. Considering actual effort in *Daily*, most subjects ended up ‘starting high and ending low’, and those who aimed to ‘start high and end low’ were the most likely to follow their planned profile of effort (Table 4). The tendency is similar for goal achievers

Figure 5: Average daily effort and goals (*Daily* and *Weekly*).



and non-achievers (Table A9). This suggests that some subjects failed to predict either that other things might get in the way, or that they might simply be less attentive to the study later in the week. For *Weekly* we do not have information on the planned profile of effort. Looking at the actual effort profile, most subjects ended up ‘starting high and ending low’ – like in *Daily* (Table A10).

Figure 5 suggests that subjects in *Weekly* worked less on Monday than subjects in *Daily*, as predicted by Hypothesis H3a. Yet, a permutation test yields  $p = 0.218$  and the effect is not robustly significant in the regressions. In addition, we do not see in *Weekly* the pattern of catching up on Friday predicted by Hypothesis H3b (Tables A13 and A14). One explanation could be that effort costs are increasing toward the end of the week. As explained above, this would weaken the prediction of a treatment difference in effort on Friday. In addition, as noted before, the patterns of daily goals suggest heterogeneity in effort costs across days. In our comparison of *Daily* with *Weekly* we can only control for the total goal but have no control for effort costs on a given weekday, which might not be balanced across treatments. Our further treatments *Aggregated* and *Daily(R)*, discussed in the next section, address this issue.

## 5 Extensions

In this section, we consider two extensions of our main treatments. The first allows us to cleanly test for effort substitution. The second explores the effects of adding a minimum work requirement in addition to goal setting.

### 5.1 Goal feedback format (*Daily(R)* vs. *Aggregated*)

By comparing *Daily* and *Weekly* we examined the overall effect of the goal format on goals and effort. Yet, as the goal level was endogenous, the control for the *total goal* in Hypothesis H1b and the corresponding regressions provided only an imperfect test of whether the difference in effort was caused by the goal bracket (daily vs. weekly) or by the difference in goals that the different treatments induced. Further, the goal setting patterns in *Daily* suggested non-constant effort costs over time, for which we could not control in our examination of effort substitution (Hypotheses H3a,b). To address these issues, we conducted treatments *Daily(R)* and *Aggregated*. We first describe the design and hypotheses.

#### 5.1.1 Design and hypotheses

**Treatments.** All subjects set daily goals like in *Daily*. Then they were randomized either to *Daily(R)* (where subjects got feedback about their daily goals, thus replicating *Daily*), or to *Aggregated* (where they got feedback about their weekly goal derived by aggregating the daily goals). There was daily feedback about goals in both treatments, as in *Daily* and *Weekly*.

**Hypotheses.** First, the treatments allow us to compare effort with daily and weekly goals ‘holding fixed’ the total goal (Hypothesis H1b). By design, the total goal should not vary across *Daily(R)* and *Aggregated*, so we can directly compare effort across treatments. Second, the treatments provide us with daily goals for both treatments that we can use as controls to obtain a cleaner test of effort substitution (H3a,b).<sup>17</sup> In addition, the treatments are of independent interest because showing the effect of exogenously shifting the framing (but

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<sup>17</sup>It is straightforward from the proofs of Propositions 2 and 5 that H3a,b continue to apply, as do H2a,b.

not the level) of goals is highly relevant to organizations (see the discussion of the study by Cadena et al. (2011) in Section 7).

### 5.1.2 Analysis

We follow the structure of Section 4, comparing across treatments the goals, effort, goal achievement, and effort patterns over time. For our hypotheses to be valid, we need to check that the data do not contradict the assumptions behind the hypotheses. In Appendix H we confirm (i) that there was no treatment difference in the goals, and (ii) that the framing of the goal feedback was successful in the sense that subjects in *Aggregated* responded to the weekly goal about which feedback was given and not to the daily goals they had initially set.

**Total effort.** We first test Hypothesis H1b. In line with the hypothesis, we find no treatment effect on total effort (permutation test,  $p = 0.247$ ; Table A12). Yet, some caution should be taken when interpreting this result as the non-significant effect might be due to the sample size.<sup>18</sup>

**Goal achievement.** In line with Hypothesis H2a, we find no treatment difference in the likelihood of achieving the total goal (Fisher’s exact test,  $p = 0.101$ ; the logit coefficient becomes insignificant once adding controls in Table A5). In line with H2b, the right panel of Figure 6 shows that more mass is concentrated just to the left of zero in *Daily(R)* than in *Aggregated*. Yet, tobit regressions of *total effort-total goal* for subjects who fell short of the total goal but attempted working ( $\text{total effort} > 0$ ) reveal no significant treatment effect (Table A6).

**Effort substitution.** Our second objective was to cleanly test for effort substitution (H3a,b) by exploiting daily goals as a control for possible heterogeneity in effort costs over the course of the week. In line with Hypothesis H3a, Figure 7 suggests that subjects in *Aggregated* worked less on Monday than subjects in *Daily*. While a permutation test has

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<sup>18</sup>Table A12 indicates that the treatment dummy is not statistically significant. Without controls, the coefficient appears large (-108.26). Given the large coefficient, there might be some concern that the result is imprecisely estimated. However, the effect size is negligible (-0.13). With controls, the coefficient is small and the effect size is close to zero.

Figure 6: Distribution of goals and effort (*Daily(R)* & *Aggregated*).

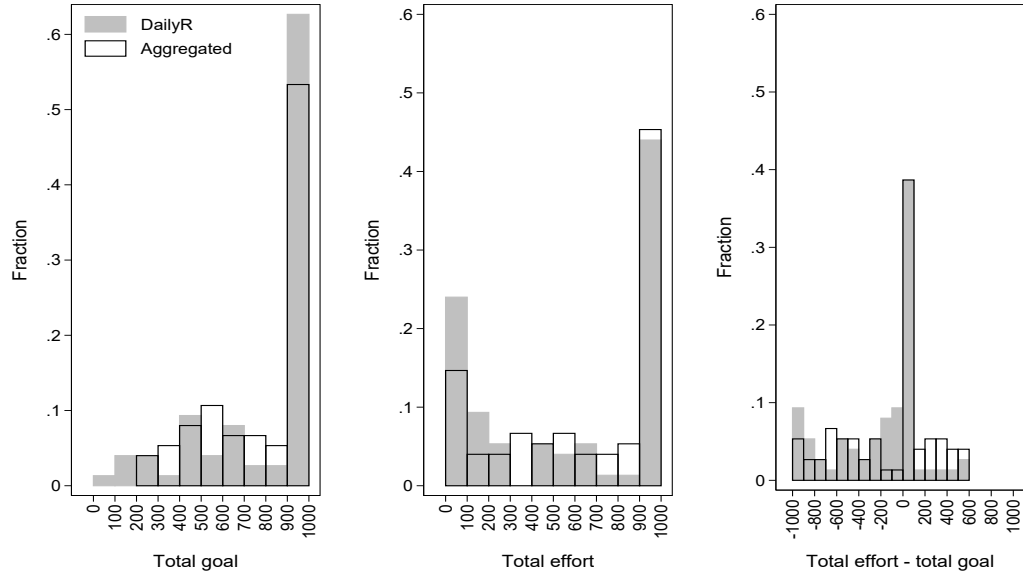
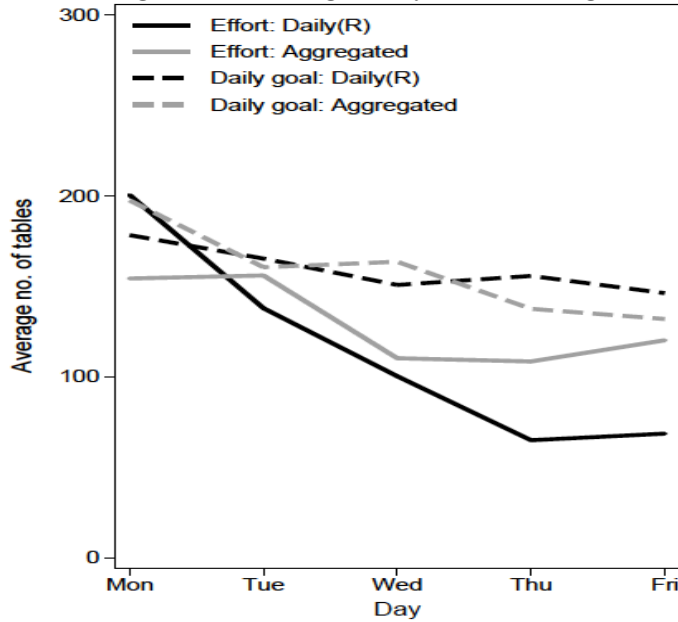


Figure 7: Average daily effort and goals.



$p = 0.154$ , the tobit coefficient becomes significant once adding controls (Table A13). Further, the figure suggests that subjects made up for the shortfall by working harder on Friday in *Aggregated* than subjects in *Daily(R)*, in line with Hypothesis H3b. This is confirmed by a permutation test ( $p = 0.017$ ) and tobit regressions (Table A14).

Table 5: Hypotheses and summary of findings

	Hypothesis	Finding	Section	Table
<b><u>Goals and effort</u></b>				
<b>H1</b>	Total goal( <i>Daily</i> )>total goal( <i>Weekly</i> )	✓	4.1	3
<b>H1b</b>	Total effort( <i>Daily</i> )>total effort( <i>Weekly</i> )	✓	4.2	3
	Controlling for total goal, total effort( <i>Daily</i> )=total effort( <i>Weekly</i> )	✓	4.2	3
<b>H1b'</b>	Total effort( <i>Daily(R)</i> )=total effort( <i>Aggregated</i> ), as total goal equal	✓	5.1	A12
<b><u>Goal achievement</u></b>				
<b>H2a</b>	Goal achievement rate( <i>Daily</i> )=goal achievement rate( <i>Weekly</i> )	✓	4.2	A5
	Goal achievement rate( <i>Daily(R)</i> )=goal achievement rate( <i>Aggregated</i> )	✓	5.1	A5
<b>H2b</b>	Goal non-achievers are closer to total goal in <i>Daily</i> than in <i>Weekly</i>	✓	4.2	A6
	Goal non-achievers are closer to total goal in <i>Daily(R)</i> than in <i>Aggregated</i>	✗	5.1	A6
<b><u>Effort substitution</u></b>				
<b>H3a</b>	Monday effort( <i>Daily</i> )>Monday effort( <i>Weekly</i> )	✗	4.4	A13
	Monday effort( <i>Daily(R)</i> )>Monday effort( <i>Aggregated</i> )	✓	5.1	A14
<b>H3b</b>	Friday effort( <i>Daily</i> )<Friday effort( <i>Weekly</i> )	✗	4.4	A14
	Friday effort( <i>Daily(R)</i> )<Friday effort( <i>Aggregated</i> )	✓	5.1	A14

Note: Tables prefaced with ‘A’ are in the online appendix.

Our treatments offer an additional way to test Hypothesis H3b. All subjects stated a daily goal for Friday. Taking this goal as a benchmark, we indeed find that subjects in *Aggregated* were more likely to work harder than that benchmark on Friday in *Aggregated* than subjects in *Daily(R)* (Fisher’s exact test,  $p = 0.009$ ; logit marginal effect 17-20 percentage points,  $p \leq 0.020$ ). We summarize our findings from this and the previous section in Table 5.

## 5.2 The impact of minimum work requirements

### 5.2.1 Is it about getting started?

Our model assumes that people ‘just’ choose their effort. In practice, the effort decision is more complex. Subjects need to follow the emailed link, start solving tables (‘get started’), and then continue to work towards their goal (‘get finished’). When we examine these two dimensions, subjects with daily goals were more likely to get started. On average there

was one extra login during the week for *Daily* vs. *Weekly* (6.7 vs. 5.5, permutation test,  $p = 0.108$ ) and subjects in *Daily* log in on more weekdays than subjects in *Weekly* (3.2 vs. 2.6,  $p = 0.014$ ). Once logged on, subjects with daily goals completed 131 tables per login in *Daily* vs. 104 in *Weekly* ( $p = 0.106$ ).

To explore further whether getting started helps subjects to get finished, we conducted treatments *DailyRequirement* and *WeeklyRequirement*. They differed from *Daily* and *Weekly*, respectively, only in that we introduced a minimal work requirement. Ariely and Wertenbroch (2002) argue that externally imposed commitment takes away flexibility and thus might be harmful – in particular for people without a self-control problem. We designed the work requirement to limit this problem. It required a subject to spend less than a minute to start working every day, but otherwise allowed for flexibility of whether and when to work. Specifically, subjects were informed that they needed to click the link in their email and complete at least one table per day to qualify for payments. If they failed to do so, they would lose all earnings for the week.<sup>19</sup>

**Predictions (without a formal theory).** The minimum work requirement does not commit subjects to fulfil a certain workload, but it commits them to ‘get started’ each day. The idea is that, once a subject clicked on the link and solved one table, he might continue to work. If subjects anticipate that the work requirement helps them to get started, thereby alleviating the problem of effort substitution, we expect them to set the same total goals in *WeeklyRequirement* and *DailyRequirement* and to provide the same total effort.

**Results.** Subjects in *DailyRequirement* aimed to complete a total of 858 tables on average, whereas those in *WeeklyRequirement* aimed for 750 (permutation test,  $p = 0.067$ , Table A3). The treatment difference is not significant in tobit regressions, but has a non-negligible effect size (Table A15). Thus, the evidence is inconclusive. Further, we observe no significant difference in effort (permutation test,  $p = 0.448$ ; Table A15), and there is no treatment difference in the number of logins ( $p = 0.278$ ), weekdays logged on ( $p = 0.195$ ), or tables completed per login ( $p = 0.717$ ). The results suggest that the motivational power of a daily

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<sup>19</sup>The only exception was if a person completed 1,000 tables before Friday – then no further actions were required (though this was not made explicit in the instructions in order to avoid an incentive to finish early).



goal comes from helping people to get started working regularly and that the work requirement does the same. The fact that only few subjects completed just the one table required (14 single-table-for-a-day logins out of 460 subject-day combinations) indeed suggests that the minimum work requirement did get subjects started working. But our analysis in the next subsection shows that there is a downside to having a work requirement.

### 5.2.2 Do work requirements complement internal commitment?

To examine whether an externally enforced minimum work requirement complements internal commitment through goals, we now compare *Daily* and *DailyRequirement* as well as *Weekly* and *WeeklyRequirement*.

**Predictions (without a formal theory).** The work requirement allows flexible distribution of effort over the five workdays. Only subjects who anticipate being away from a computer with internet access for a whole day – a situation that is quite unlikely during term time – have a reason to drop out. So the work requirement should only lead to a small increase in drop out. Anticipating that the work requirement gets them to work more regularly, subjects should set a higher total goal and then provide a higher effort in *DailyRequirement* (*WeeklyRequirement*) than in *Daily* (*Weekly*).

**Results.** Goals did not differ between *Daily* and *DailyRequirement* or between *Weekly* and *WeeklyRequirement* (permutation tests,  $p = 0.172$  and  $p = 0.262$ ; Table 6), but the effect sizes are non-negligible. We observe *lower* effort in *DailyRequirement* than in *Daily*, and no significant effort difference between *WeeklyRequirement* and *Weekly* (permutation tests,  $p = 0.008$  and  $p = 0.616$ ; Table 6).

To understand the mechanisms behind the surprising result that the work requirement harmed performance, we perform an exploratory analysis of the dropout behavior in the different treatments. The data contradict our premise that the work requirement does not affect drop-out. Table 7 shows that the dropout probability increased from around 4 percent in *Daily* and *Weekly* to over 30 percent in *DailyRequirement* and *WeeklyRequirement*. Dropout occurred mostly right from the start and was not caused by subjects starting with the task to then stop later during the week.

Table 6: Impact of work requirement on total goal and on total effort.

Dependent variable Sample <sup>a</sup>	Total goal	Total effort (tables counted)			
	All	All		No dropout	
	(1)	(2)	(3)	(4)	(5)
DailyRequirement <sup>b</sup> vs. Daily	118.91 (100.53)	-323.35** (154.97)	-371.75** (150.40)	151.83 (142.91)	80.33 (130.22)
Margin.effect(requirement) <sup>c</sup>	51.34	-160.07**	-193.16**	74.03	44.09
Effect size <sup>d</sup>	0.17	-0.43	-0.52	0.21	0.13
WeeklyRequirement <sup>b</sup> vs. Weekly	144.70 (112.58)	13.30 (158.89)	-43.27 (153.12)	496.37*** (138.57)	410.70*** (125.79)
Margin.effect <sup>c</sup>	77.12	6.79	-22.75	257.27***	220.77***
Effect size <sup>d</sup>	0.23	0.02	-0.06	0.67	0.57
Control for total goal	–	no	yes	no	yes
N	247	247	247	209	209

Notes. Each cell reports the tobit coefficient (with robust standard error in parenthesis) from a separate regression with the omitted category given by “vs.”. Specifications include baseline productivity, a gender dummy, and a constant. Results are similar without controls or with the full set of controls. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>All: all subjects in *Daily*, *Weekly*, *DailyRequirement*, *WeeklyRequirement*. No dropout: subsample with total effort > 0/satisfying daily login requirement. <sup>b</sup>Dummy for *DailyRequirement* or *WeeklyRequirement*, respectively. <sup>c</sup>Tobit marginal effect on the censored latent variable of the minimum work requirement vs. no requirement (at the means of control variables).

<sup>d</sup> $\frac{\text{Margin.effect}}{\text{Standard deviation of outcome in } \textit{Daily} \text{ or } \textit{Weekly}, \text{ respectively}}$ .

Table 7: Dropout

Treatment	Drop out/zero effort <sup>a</sup>		Login Monday		Total effort (mean)	
	N	Percent	N	Percent	Full sample	Not dropped out
Daily	78	3.8	62	79.5	690.0	717.6
Weekly	77	3.9	51	66.2	520.8	541.9
DailyRequirement	47	36.2	35	74.5	487.1	744.9
WeeklyRequirement	45	33.3	32	71.1	558.3	811.1
Marginal effect of minimum work requirement vs. no requirement (percentage points). <sup>b</sup>						
	Drop out/zero effort <sup>a</sup>			Login Monday		
DailyRequirement vs. Daily <sup>c</sup>	30.13*** (7.53)			-5.71 (8.12)		
WeeklyRequirement vs. Weekly <sup>c</sup>	29.10*** (7.58)			4.26 (8.67)		

Notes. <sup>a</sup>Dropout: counted zero tables/failed to satisfy the daily login requirement. <sup>b</sup>Each cell reports the logit marginal effect (with robust standard error in parenthesis) from a separate regression with the omitted category given by “vs.”. Specifications include total goal, baseline productivity, a gender dummy, and a constant. Results are similar without controls or with the full set of controls. Sample: *Daily*, *Weekly*, *DailyRequirement*, *WeeklyRequirement*, *N* = 247. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>c</sup>Dummy for *DailyRequirement* or *WeeklyRequirement*, respectively.

Further, there was no significant effect of the requirement on the likelihood of working on Monday (bottom of Table 7). Thus, the work requirement appears to have pushed subjects to drop out who otherwise would have started working later than Monday – instead of getting them to start earlier.

In a next step, we examine whether the work requirement had any benefits for the subjects who did participate. Conditional on participation, the requirement did not have a significant effect on the total goal (*DailyRequirement* vs. *Daily* tobit coefficient 113.43,  $se = 116.63$ ,  $p = 0.332$ ; *WeeklyRequirement* vs. *Weekly* 157.31,  $se = 131.72$ ,  $p = 0.234$ ;  $N = 209$ ). For effort, Table 6 (‘No dropout’ columns) reveals no significant difference between *DailyRequirement* and *Daily*, but significantly higher effort in *WeeklyRequirement* than in *Weekly*. If we consider each workday separately, subjects in *WeeklyRequirement* completed significantly more tables on any given day than subjects in *Weekly*.<sup>20</sup> That is, conditional on participation, subjects worked more steadily in *WeeklyRequirement* than in *Weekly*.

Taken together, these results suggest that a daily goal without an additional, externally enforced work requirement leads to the highest performance. The narrow goal bracket already motivates individuals to ‘get started’. At the same time, it avoids the problem of dropout that the work requirement causes. When facing a weak internal commitment device (a weekly goal), the work requirement however does seem to benefit individuals who do not drop out, helping them to ‘get started’.

## 6 Discussion

In this section, we discuss (alternative) mechanisms and extensions of our experiment.

### 6.1 Alternative mechanisms and explanations

**Goal non-achievement.** Almost half of the subjects failed to reach their goal (Table 2). This is hard to explain with naïveté about the present bias alone, because our model predicts non-achievement only for a relatively severe present bias (cf. Figure 2). Mistakes in

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<sup>20</sup>Mon: tobit coefficient 96.76\*\*,  $se = 48.44$ ; Tue: 71.06\*\*,  $se = 34.42$ ; Wed: 186.32\*\*\*,  $se = 43.40$ ; Thu: 109.35\*\*,  $se = 43.16$ ; Fr: 108.84\*\*,  $se = 44.28$ ;  $N = 104$ . Results are similar without controlling for total goal.

updating beliefs might be one cause for not reaching a weekly goal. In our model, partially naïve individuals hold wrong beliefs about their effort. Adjusting for past observed effort, they keep updating their beliefs about future effort and goal achievement. Yet, such updating is not necessary under daily goals. Thus, the observed non-difference in goal achievement between daily and weekly goals speaks against mistakes in updating being a major source of goal non-achievement.

Further, it seems unlikely that non-achievement is due to a lack of time. 97 percent of subjects in *Daily*, *Weekly*, *Daily(R)*, and *Aggregated* set a goal that they could achieve in the maximum number of hours they could devote to the task (reported before setting goals), when taking into account their baseline productivity. Subjects also reported the maximum number of tables they thought they could solve in the time they had available. By this measure, 67 percent of goals were realistic. There is mixed evidence whether those who set such a ‘subjectively realistic goal’ were more likely to achieve their goal. There is no significant effect (Fisher’s exact test,  $p = 0.145$ ) but the logit marginal effect of 9.44 percentage points is non-negligible ( $p = 0.134$ ,  $N = 305$ ). Even so, still only 58 percent of subjects with a ‘subjectively realistic goal’ achieved their goal.

The descriptive evidence in Section 4.4 suggests that many subjects may have failed to correctly predict changes in circumstances later in the week. Irrespective of their original plans, subjects in *Daily* tended to provide lower effort at the end of the week than at the beginning of the week (Table 4). Some subjects might have underestimated how annoying the task was and decreased their effort after learning this. Or they might have failed to predict an increase in marginal costs of effort over the course of the week (see Section 4.4). For example, we observe that goal achievers had a lower total goal than non-achievers (permutation test,  $p = 0.089$ , but not significant in tobit regressions,  $p \geq 0.133$ ). That is, achievers possibly better predicted the increase in marginal costs. Theoretically, this would have an effect similar to that of naïveté about the present bias.<sup>21</sup> Another reason could simply be that attention to the study decreased, the longer the study carried on. In line with this explanation, the share of subjects with at least one login on a given day decreased in both main treatments from Monday to Friday (among those who had not yet reached

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<sup>21</sup>One can easily see this in our parametric example with quadratic effort costs ( $c_t(e_t) = (c_t e_t^2)/2$ ), where  $c_t$  has an inversely proportional effect to  $\beta$  on  $e_{max,t}(\beta) = (2\beta)/c_t$ .

their total goal before Friday): from 75 to 59 percent in *Daily* and from 51 to 37 percent in *Weekly*.

**Mistakes in goal setting.** Mistakes could explain our findings if they had led subjects to set a higher total goal when they picked five daily goals instead of one weekly goal. First, subjects might have felt reluctant to choose a large number, which would have led to a lower total goal in *Weekly* compared to *Daily*. If reluctance of picking a large number was a driver, the proportion of subjects with  $total\ goal = 1000$  should be lower in *Weekly* than in *Daily*. This is not the case (Fisher’s exact test,  $p = 0.631$ ; logit regressions  $p \geq 0.445$ ). Second, subjects may have made small, upward biased mistakes when setting goals. As subjects had to make five choices when setting daily goals compared to one when setting a weekly goal, the sum of small mistakes would have been larger when setting daily goals. This explanation would predict a higher likelihood of the total goal being ‘too high’ in *Daily* compared to *Weekly*. A proxy is the estimated maximum number of tables that subjects thought they would be able to complete if they used all the spare time they had available to work on the tasks – a question they answered right before learning that they should set goals. Comparing the proportion of subjects for whom the total goal exceeds this number (30 percent in *Weekly* vs. 29 percent in *Daily*), we reject a treatment difference in the proportion of ‘too high’ goals (Fisher’s exact test,  $p = 1.000$ ; logit regressions  $p \geq 0.913$ ).

**Experience with the task.** A random effects panel regression yields no meaningful effect of experience on productivity (0.6 seconds *longer* completion time for the 1000<sup>th</sup> table compared to the first table). To test for possible effects of experience on goal setting and achievement, we exploit variation in how we recruited subjects (see Appendix E). We compare 71 subjects in *Daily* (out of 78), who had 2-3 hours experience with the real-effort task from a prior online experiment (Epper et al., 2018), with subjects in *Daily(R)*, who had no experience before. We find no significant difference in total goal (permutation test  $p = 0.547$ ; tobit regressions  $p \geq 0.228$ ) or in the likelihood of achieving the goal (Fisher’s exact test,  $p = 0.382$ ; logit regressions,  $p \geq 0.242$ ).

**Wrong beliefs about effort costs.** Suppose self 0 thinks that effort costs are  $\underline{c}(e)$ , while on the first working day they turn out to be  $\bar{c}(e) > \underline{c}(e)$ . With daily goals, self 0 sets goals

$g_t = e_0^*$ , where  $b'(e_0^*) = \underline{c}'(e_0^*)$ . This goal however might not be implementable (if  $\bar{e}_{max} < e_0^*$ , where  $\beta(1 + b'(\bar{e}_{max})) = \bar{c}'(\bar{e}_{max})$ ). In this case, both with daily and weekly goals, all selves would provide  $\bar{e}_{max}$  each period and fail to achieve the total goal. If  $\bar{e}_{max} > e_0^*$ , then all selves will stick to their daily goals. Yet, under a weekly goal it can happen that the originally planned  $\hat{e}_{T,0} > \bar{e}_{max} > e_0^*$ , so that self  $T$  will deviate and provide  $\bar{e}_{max}$ . Anticipating this, previous selves will increase their effort so that effort substitution will be less pronounced. The individual may or may not achieve the goal. Thus, wrong beliefs about effort costs could explain some of the observed goal non-achievement.

**Uncertainty about effort costs.** Uncertainty about effort costs translates into uncertainty about effort. In a theoretical model, Koch and Nafziger (2016) demonstrate the effect of uncertainty on narrowly or broadly bracketed goals. A broad bracket allows pooling of risks across tasks, so that the individual suffers less often a loss due to goal non-achievement. Holding effort fixed, this increases the utility of the individual. Yet, exactly this risk pooling effect might dampen incentives because it is the fear of not achieving the goal that makes individuals strive for their goal. If there is little uncertainty, the ability to implement a better decision under narrow bracketing trumps the benefits from risk pooling under broad bracketing. Overall, this suggests that uncertainty will tend to strengthen the prediction of a lower total goal and lower total effort in *Weekly* compared to *Daily*.

Now consider learning about effort costs. Learning strategies (like “start small to learn your costs”) can take place in the same way under daily and weekly goals and thus would not lead to treatment differences in the absence of a self-control problem. Subjects may revise their goals after revelation of uncertainty, failing to meet expected effort (due to partial naïveté), or learning about effort costs might. Realizing that goals cannot be reached or are very expensive to reach, the subject revises them downward. As Koch and Nafziger (2016) demonstrate, goals are still effective in this case, but potentially have a lower motivational power so that the initial goal is not achieved. Yet, effort substitution also occurs when allowing for goal revision. Thus, our main predictions regarding treatment differences between *Daily* and *Weekly* carry over. In any case, our data provide no evidence of experience affecting goal setting or goal achievement (see above).

**Time horizon.** The time frame of the project does not matter for the theoretical predictions. Our model predicts for any  $T > 1$  that a broad goal leads to effort substitution and lower effort compared to narrow goals. This prediction however relies on the assumption that the next period is in ‘the future’, i.e., that self  $t$  discounts by the present bias  $\beta$  the payoffs occurring in  $t+1$  and beyond. Studies on time discounting typically consider periods of one day (e.g. Laury et al., 2012), but one study shows evidence of present bias even over a 5-minute interval (McClure et al., 2007).

In our experiment, we opted for a one-week time horizon and chose to remind people about their goals. Our empirical findings thus seem most applicable to settings where individuals face salient goals over a relatively short time horizon. Examples are project work or studying for an exam. With longer time horizons, goals may be less salient and reminders may be more important. An advantage of daily goals could be that they act as reminders by themselves. In addition, the bracketing of goals may affect how easy it is to monitor goal progress. Take for example a cab driver who can choose his working hours. It may be easier to monitor a fixed daily goal (such as “work 8 hours” or “earn \$150”) than to monitor progress toward a longer term goal because the latter requires keeping an account for accumulated effort.

**Dropout.** The literature on *hidden costs of control* (Falk and Kosfeld, 2006) suggests aversion to being controlled as a possible explanation for dropout in the treatments with the work requirement.<sup>22</sup> Given the evidence that externally imposed goals increase performance (see, among others, Goerg and Kube, 2012), dropout may have been a response specifically to having been forced to work every day and not so much a response to having been forced to set goals. In particular, subjects might have reacted to having been *forced* to count *one* table. This very unambitious target was perhaps perceived as a nuisance rather than as an encouragement. Other external requirements or incentive schemes may work better. The optimal design of such schemes is an avenue for future research.

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<sup>22</sup>Another explanation builds on Bénabou and Tirole (2004). Here external control crowds out intrinsic motivation because it is a bad signal about effort costs. We cannot test this explanation. However, we believe that intrinsic motivation for the counting task is likely to have been low from the start.



**Power.** We performed power calculations only ex-post. Such ex-post calculations should be taken with care. We provide them to give some rough feeling for the sample size. Figure A4 in the online appendix shows the power analysis for the main Hypothesis H2 that we test. Based on the observed treatment difference in effort of 169 tables for *Daily* vs. *Weekly* and the actual number of participants in each treatment, the ex-post calculated power of a two-sided test would be 0.54, 0.77, and 0.85, respectively, for a significance level of 0.01, 0.05, and 0.1. That is, depending on the significance level, there would be a 54 to 85 percent chance that we fail to reject the null hypothesis of no treatment difference even if, in fact, there was a treatment difference. Thus, our main study would be underpowered for  $\alpha \leq 0.05$ , while for  $\alpha = 0.1$  it would be correctly powered. For *Daily* vs. *Aggregated* we have a similar sample size. Comparisons involving treatments *WeeklyRequirement* and *DailyRequirement* have lower sample sizes and hence are most likely underpowered for conventional significance levels. On account of the limited power of tests, as remarked in the text, some care should be taken when interpreting statistically insignificant findings that however show non-negligible effect sizes (reported in the tables).

## 6.2 Possible extensions

**Endogenous goal bracket.** We exogenously assigned the goal bracket (daily or weekly) to identify causal effects of the goal bracket. A direction for future research would be to let subjects choose their goal bracket. While such a treatment does not identify causal effects, it may reveal interesting correlations between the choice of the bracket, effort choices, and individual characteristics. Koch and Nafziger (2019) report some survey evidence in this direction.

**Goal revision.** As discussed above, one reason for goal non-achievement may have been that subjects revised their goals. Future research could provide subjects with an explicit opportunity to revise their goals. This opens-up for new questions on the design of goals. Is the possibility of goal revision good or bad for overcoming self-control problems? While goal revision weakens the motivational power of goals (Koch and Nafziger, 2016), it allows subjects to react to resolution of uncertainty and thereby make better choices. For example, van Lent (2018) gave students (either expectedly or unexpectedly) the opportunity to revise

their goals. He observes no difference in grades between treatments with goal revision and without. Future research could control the type of uncertainty that subjects face, study the exact effort patterns (for daily vs. weekly goals), or study whether goal revision can explain goal non-achievement.

**Day-by-day goal setting.** A further question is how the front-end delay in goal setting matters. In our study, subjects in *Daily* had to think about an entire profile of goals for the next week. Yet, some people might rather focus on their goals one day at a time, i.e., set their goal for Tuesday after having concluded work on Monday, and so on. Theoretically, as long as no uncertainty is resolved, it does not matter whether the individual sets the entire goal profile in advance or sets a goal day-by-day. Allowing subjects to set goals each day in *Daily* however would imply that we do not only vary the goal bracket in comparison to *Weekly*, but also allow subjects to react to possible resolution of uncertainty or to a better understanding of their present bias. It thus seems likely that subjects in *Daily* would perform better than in *Weekly*.

**Disaggregation of weekly goals.** Another potential effect of goal bracketing is that achieving a weekly goal is cognitively more challenging than achieving a daily goal. Individuals have to understand how much effort to put in each day to achieve the weekly goal. Errors could lead both to under- or overshooting of the goal. To examine such channels, one could let subjects set a weekly goal and then disaggregate the goal into daily goals for them. Designing such a treatment however poses some challenges as to how exactly to disaggregate the weekly goal.

## 7 Conclusion

We provide a theoretical framework and experimental evidence for the motivational benefits of daily goals. In an online experiment, we exogenously assigned the goal bracket (daily or weekly) and let subjects choose their goals. Subjects worked harder under daily goals than under a weekly goal. The increase in effort was primarily related to the higher total goal level with daily goals compared to a weekly goal. In additional treatments, we exogenously

shifted the framing (but not the level) of goals. Here subjects were less likely to procrastinate effort to the end of the week when we reminded them about their daily goals than when we reminded them, with the same frequency, about their aggregated weekly goal.

We conclude by discussing some broader implications of our findings. Many organizations struggle with a suboptimal effort allocation of their employees over time, often manifested in spikes of effort immediately prior to a bonus deadline (e.g. Asch, 1990). The effort substitution over time predicted by our model and found in the experiment is consistent with these spikes. Employees facing monthly performance targets may naturally adopt a monthly goal bracket for themselves. Our findings suggest that reframing a bonus threshold in terms of smaller and more frequent narrow goals can have beneficial effects.

In line with this, evidence from a field experiment that Cadena et al. (2011) ran at a Columbian bank shows how more narrowly bracketed goals led to better effort allocation over time. The bank faced the problem that its loan officers concentrated their effort for sourcing new clients and credit collection at the end of each month, just before monthly bonuses were calculated. This increased the costs of cash flow management for the bank and it contributed to loan officers feeling stressed during the second half of the month. In their intervention, Cadena et al. (2011) increased the frequency with which employees received reminders about goals from monthly to weekly periodicity without substantially altering the bonus structure. This led to an 18 percent (10 percent) increase in new loans (renewal of loans) on average in the first two weeks of each month, without significantly affecting the overall level of loans per month or the quality of loan portfolios. In addition, workers reported lower stress levels.

Our findings tie in, more generally, with evidence on motivational benefits from externally inducing subjects to bracket narrowly and suggests ways to design interventions that can help individuals who struggle with self-control problems. For example, Soman and Cheema (2011) assigned a savings goal to Indian laborers. In the baseline condition, workers received their weekly wage in one envelope. In the treatment condition, the wage payments were split across two envelopes, with one containing the amount that the worker expressed as a savings goal. Actual savings in the treatment condition were higher than those in the baseline condition. Soman and Gourville (2001) found that single-performance ticket holders were more likely to carry through with the ex-ante desirable choice of seeing a theatre performance

than multi-performance ticket holders. Gourville and Soman (2002) observe that members who self-selected into paying their gym membership fee each month attended the gym more regularly than those who paid the same overall fee annually, semi-annually, or quarterly. Relatedly, daily repayments of microcredit loans are often observed in developing countries (e.g. Afzal et al., 2017). Bauer et al. (2012) argue that such daily repayments act as external commitment devices for individuals with a self-control problem.

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# Online Appendix for “Motivational Goal Bracketing: An Experiment”

by Alexander K. Koch and Julia Nafziger

## A Proofs

### A.1 Proof of Proposition 1 (including the case of partial naïvité)

The proof covers also the case of partial naïvité discussed in Section 3.3. The maximal implementable goals  $e_{max}(\beta)$  and  $e_{max}(\hat{\beta})$  are defined by (4) by plugging in  $\beta$  and  $\hat{\beta}$ , respectively. The preferred effort of self 0,  $e_0^*$ , is characterized by (1). Note that  $\frac{de_{max}(\beta)}{d\beta} = -\frac{b'(e_{max})+1}{\beta b''(e_{max})-c'(e_{max})} > 0$ . So  $\hat{\beta} > \beta$  implies  $e_{max}(\hat{\beta}) > e_{max}(\beta)$ . Further, (1) and (4) imply that  $e_{max}(\beta) > e_0^*$  for  $\beta = 1$  and  $e_{max}(\beta) < e_0^*$  for  $\beta = 0$ . By the intermediate value theorem, there exists a unique  $\check{\beta} \in (0, 1)$  such that  $e_{max}(\beta) \geq e_0^*$  for all  $\beta \geq \check{\beta}$  and  $e_{max}(\beta) < e_0^*$  for all  $\beta < \check{\beta}$ .

For  $e_{max}(\hat{\beta}) \geq e_0^*$  the utility of self 0 is maximized with  $g_t^* = e_0^*$ . A sophisticated individual ( $\hat{\beta} = \beta$ ) always achieves his goals. A partially naïve individual ( $\hat{\beta} > \beta$ ) only achieves his goals if, in addition,  $e_{max}(\beta) \geq e_0^*$ . Otherwise he falls short of the goal and only provides  $e_{max}(\beta)$  in each period.

For  $e_{max}(\hat{\beta}) < e_0^*$ , the utility of self 0 is maximized with  $g_t^* = e_{max}(\hat{\beta})$ . A sophisticated individual always achieves his goals, while a partially naïve individual always fails his goals and provides  $e_t = e_{max}(\beta) < e_{max}(\hat{\beta})$  in each period. (2) and (4) imply that  $e_{max}(\beta) > e_t^*$ .

### A.2 Maximal implementable effort under a weekly goal

**Lemma 1** *No matter whether  $\hat{\beta} = \beta$  or  $\hat{\beta} > \beta$ , the maximal implementable effort in a given period with a weekly goal  $G$  is the same as that for daily goals:  $e_{max}(\beta)$  defined in (4). Let  $\mathcal{A}_t$  denote the set of effort levels  $e_t \leq e_{max}(\beta)$  such that self  $t \in \{1, \dots, T\}$  believes that  $G$  still will be achieved. If  $\mathcal{A}_t \neq \emptyset$ ,  $e_t \in \mathcal{A}_t$ . If  $\mathcal{A}_t = \emptyset$ ,  $e_t = e_{max}(\beta)$ .*

**Proof.**

Consider a weekly goal  $G$ . The utility of self  $\tau$  for an effort level  $e_\tau$  and anticipated effort

responses  $\hat{e}_{t,\tau}$  for  $t > \tau$  that overall lead to failing  $G$  is

$$\beta b(e_t) - c(e_t) + \beta \sum_{t=\tau+1}^T (b(\hat{e}_{t,\tau}) - c(\hat{e}_{t,\tau})) - \beta \left( e_\tau + \sum_{t=\tau+1}^T \hat{e}_{t,\tau} + \sum_{t=1}^{\tau-1} e_t - G \right). \quad (5)$$

Taking the derivative w.r.t.  $e_t$ , (5) is increasing for  $e_t \leq e_{max,\tau}^b(\beta)$ , defined by

$$\beta b'(e_{max,\tau}) - c'(e_{max,\tau}) + \beta = \sum_{t=\tau+1}^T \left[ \beta (b'(\hat{e}_{t,\tau}) - c'(\hat{e}_{t,\tau})) \left( -\frac{d\hat{e}_{t,\tau}}{de_\tau} \right) \right], \quad (6)$$

That is, self  $\tau$  will provide up to  $e_{max,\tau}^b(\beta)$ . We now show that  $e_{max,\tau}^b(\beta) = e_{max}(\beta)$  for all  $\tau$ . Solving backward, the derivative of the utility of self  $T$  for effort levels  $e_T$  at which the goal is not yet reached (i.e.,  $e_T < G - \sum_{t=1}^{T-1} e_t$ ) is  $\beta [b'(e_T) + 1] - c'(e_T)$ . It is strictly positive for  $e_T < e_{max}(\beta)$  and strictly negative for  $e_T > e_{max}(\beta)$ , where  $e_{max}(\beta)$  is defined in (4). That is, self  $T$  responds to a goal not yet reached (e.g. because of deviations by previous selves from the effort pattern anticipated by self 0) with  $e_T = \min\{G - \sum_{t=1}^{T-1} e_t, e_{max}(\beta)\}$ . Thus,  $e_{max,T}^b(\beta) = e_{max}(\beta)$ . Now consider self  $T-1$ . He believes that self  $T$  will increase effort up to  $e_{max}(\hat{\beta})$  to prevent failing the goal  $G$ . At any effort level  $e_{T-1}$  where the  $G$  would be failed even with the maximum anticipated effort response of self  $T$  (i.e.,  $e_{T-1} < G - \sum_{t=1}^{T-1} e_t - e_{max}(\hat{\beta})$ ) we have that  $\frac{d\hat{e}_{T,T-1}}{de_{T-1}} = 0$ . Because the right-hand side of (6) then is zero, it reduces to (4) and  $e_{max,T-1}^b(\beta) = e_{max}(\beta)$ . Iterating backwards, we get for all previous periods  $\tau$  that  $e_{max,\tau}^b(\beta) = e_{max}(\beta)$ . The last part follows from (5) being increasing for  $e_t \leq e_{max}(\beta)$ . ■

### A.3 Goal achievement and effort patterns for a sophisticated individual facing a weekly goal

We start by proving some intermediate results on goal achievement and effort patterns for a sophisticated individual ( $\hat{\beta} = \beta$ ) that will be used in the proofs of Propositions 2, 3, and 5. For a partially naïve individual ( $\hat{\beta} > \beta$ ), the results will pin down the beliefs of self 0 about future effort.

#### A.3.1 Goal achievement

Self 0 wants his future selves to put in more effort than  $e_t^*$ , which is the effort in a given period  $t$  preferred by self  $t$  in the absence of comparison utility (defined in (2)). We first

show that the individual achieves any weekly goal  $G \in [T e_1^*, T e_{max}(\beta)]$  (Lemma 2). We then characterize the effort patterns that the individual will choose for such goals (Lemmas 4 and 5).

**Lemma 2** *Suppose  $\hat{\beta} = \beta$ . The individual achieves any weekly goal  $G \in [T e_1^*, T e_{max}(\beta)]$ , i.e.,  $\sum_{t=1}^T e_t = G$ .*

**Proof.**

Recall that (2) and (4) imply  $e_{max}(\beta) > e_1^*$ . Suppose that reaching the goal  $G$  still is feasible in period  $t > 0$ , i.e.,  $G - \sum_{\tau=1}^{t-1} e_\tau \leq (T - t + 1) e_{max}(\beta)$ . It is never optimal for self  $t$  to choose some effort level  $e_t < e_{max}(\beta)$  for which future selves will not make up for the shortfall (Lemma 1). Hence, self  $t$  chooses  $e_t$  so that  $G$  will either be achieved or overachieved. Overachievement is never optimal from the perspective of any self. This is trivial for  $t = T$ , For  $t < T$ , future selves would lower their effort in response to any effort by self  $t$  that would lead to overachievement. By way of contradiction, suppose that self  $t$  anticipates that the goal  $G$  will be overachieved in the end. His optimal effort then satisfies  $\beta b'(e_t) = c'(e_t)$ , which is independent of future effort and therefore coincides with (2). That is, he would chose  $e_t = e_1^*$ . But as  $T e_1^* \leq G$  this effort profile cannot lead to overachievement, leading to a contradiction. Thus,  $G$  will be exactly achieved. ■

As a direct corollary to Lemmas 1 and 2, we obtain:

**Lemma 3** *Suppose  $\hat{\beta} = \beta$ . With a weekly goal  $G \geq T e_{max}(\beta)$ , the individual will provide  $e_{max}(\beta)$  in each period.*

### A.3.2 Effort profile over time

Our next result shows that a reduction in effort by self  $\tau$  leads future selves  $t > \tau$  to increase effort, unless constrained by  $e_{max}(\beta)$ .

**Lemma 4** *Suppose  $\hat{\beta} = \beta$  and  $T e_1^* < G < T e_{max}(\beta)$ . Then,  $\frac{de_t}{de_\tau} \leq 0$  for all  $t > \tau$ , with strict inequality if  $e_t < e_{max}(\beta)$ .*

**Proof.**

The first-order condition for self  $t < T$  for an interior solution  $e_t < e_{max}(\beta)$  is

$$\beta b'(e_t) - c'(e_t) = \sum_{k=t+1}^T \left[ \beta (b'(e_k) - c'(e_k)) \left( -\frac{de_k}{de_t} \right) \right]. \quad (7)$$

From the implicit function theorem we get:

$$\frac{de_t}{de_\tau} = \sum_{k=t+1}^T \left[ \underbrace{\frac{\beta [b''(e_k) - c''(e_k)] \left( -\frac{de_k}{de_t} \right)}{\beta b''(e_t) - c''(e_t)}}_{\equiv \phi(e_k, e_t)} \left( \frac{de_k}{de_\tau} \right) \right] \quad \text{for } \tau < t. \quad (8)$$

Note that  $\beta [b''(e_k) - c''(e_k)] / [\beta b''(e_t) - c''(e_t)] > 0$  because  $b''(\cdot) \leq 0$  and  $c''(\cdot) > 0$ . Moreover, Lemma 2 implies that

$$\sum_{\tau=t+1}^T \frac{de_\tau}{de_t} = -1. \quad (9)$$

We now show that for  $e_t < e_{max}(\beta)$  we have  $\frac{de_t}{de_\tau} < 0$  for all  $t, \tau \in \{1, \dots, T\}$ ,  $\tau < t$ .

**Step 1.** Note that  $\frac{de_T}{de_{T-1}} = -1$  implies  $\phi(e_T, e_{T-1}) > 0$ . Setting  $t = T - 1$  in (8) gives

$$\frac{de_{T-1}}{de_\tau} = \phi(e_T, e_{T-1}) \left( \frac{de_T}{de_\tau} \right) \quad \text{for } \tau < T - 1,$$

which therefore implies that

$$\text{sign} \left( \frac{de_{T-1}}{de_\tau} \right) = \text{sign} \left( \frac{de_T}{de_\tau} \right) \quad \text{for } \tau < T - 1. \quad (10)$$

Setting  $t = T - 2$  in (9), we have

$$\frac{de_{T-1}}{de_{T-2}} + \frac{de_T}{de_{T-2}} = -1. \quad (11)$$

Together, (10) and (11) imply that  $\frac{de_t}{de_{T-2}} < 0$  for  $t > T - 2$ .

**Step 2.** Setting  $t = T - 2$  in (8) gives

$$\frac{de_{T-2}}{de_\tau} = \underbrace{\phi(e_{T-1}, e_{T-2}) \left( \frac{de_{T-1}}{de_\tau} \right) + \phi(e_T, e_{T-2}) \left( \frac{de_T}{de_\tau} \right)}_{\text{same sign (by (10))}} \quad \text{for } \tau < T - 2. \quad (12)$$

From step 1 we know that  $\phi(e_{T-1}, e_{T-2}) > 0$  and  $\phi(e_T, e_{T-2}) > 0$ , which together with (12) imply that

$$\text{sign} \left( \frac{de_{T-2}}{de_\tau} \right) = \text{sign} \left( \frac{de_{T-1}}{de_\tau} \right) = \text{sign} \left( \frac{de_T}{de_\tau} \right) \quad \text{for } \tau < T - 2.$$

Setting  $t = T - 3$  in (9),

$$\frac{de_{T-2}}{de_{T-3}} + \frac{de_{T-1}}{de_{T-3}} + \frac{de_T}{de_{T-3}} = -1,$$

and using that all terms on the left-hand side have the same sign, we get  $\frac{de_t}{de_{T-3}} < 0$  for  $t > T - 3$ .

**Step 3.** Continuing the iteration, plugging  $t < T$  into (8), we obtain

$$\frac{de_t}{de_\tau} = \sum_{k=t+1}^T \underbrace{\phi(e_k, e_t)}_{> 0} \underbrace{\left( \frac{de_k}{de_\tau} \right)}_{\text{same sign}} \quad \text{for } \tau < t.$$

(by prev. step) (by prev. step)

This means that all terms on the left-hand side of (9) have the same sign, which implies that  $\frac{de_t}{de_\tau} < 0$  for  $t > \tau$ . By extension, including corner solutions  $e_t = e_{max}(\beta)$ , we have  $\frac{de_t}{de_\tau} \leq 0$ .

■

**Lemma 5** Suppose  $\hat{\beta} = \beta$  and  $T e_1^* < G < T e_{max}(\beta)$ . Effort is (weakly) increasing over time:  $e_1 < e_2 \leq e_3 \leq \dots \leq e_T \leq e_{max}(\beta)$ .

**Proof.**

It follows from Lemma 2 that each self  $t < T$  will choose  $e_t$  such that  $G \leq e_t + (T - t) e_{max}(\beta) + \sum_{\tau=1}^{t-1} e_\tau$ . We solve backward. Anticipating the behavior of self  $T$ , self  $T - 1$  maximizes

$$\beta b(e_{T-1}) - c(e_{T-1}) + \beta \left[ b \left( \underbrace{G - e_{T-1} - \sum_{t=1}^{T-2} e_t}_{=e_T} \right) - c \left( G - e_{T-1} - \sum_{t=1}^{T-2} e_t \right) \right].$$

Using (9) yields the first-order condition for an interior solution  $e_{T-1} < e_{max}(\beta)$ :

$$\beta b'(e_{T-1}) - c'(e_{T-1}) = \beta [b'(e_T) - c'(e_T)]. \quad (13)$$

As  $\beta b'(e) - c'(e) < \beta [b'(e) - c'(e)]$ , it follows that either  $e_{T-1} < e_T$  or we have a corner solution  $e_{T-1} = e_T = e_{max}(\beta)$ . Similarly, the first-order condition of self  $T - 2$  for an interior solution  $e_{T-2} < e_{max}(\beta)$  is given by:

$$\begin{aligned} & \beta b'(e_{T-2}) - c'(e_{T-2}) \\ &= \beta [b'(e_T) - c'(e_T)] \left( -\frac{de_T}{de_{T-2}} \right) + \beta [b'(e_{T-1}) - c'(e_{T-1})] \left( -\frac{de_{T-1}}{de_{T-2}} \right). \end{aligned} \quad (14)$$

Lemma 2 implies that  $\frac{de_{T-1}}{de_{T-2}} + \frac{de_T}{de_{T-2}} = -1$ . Hence, if  $e_{T-1} = e_T = e_{max}(\beta)$ , then  $\beta b'(e) - c'(e) < \beta (b'(e) - c'(e))$  implies that  $e_{T-2} < e_{T-1}$ , or we have a corner solution  $e_{T-2} = e_{T-1} = e_T = e_{max}(\beta)$ . Now if we have an interior solution  $e_{T-1} < e_{max}(\beta)$ , we can substitute from (13)

$$\begin{aligned} & \beta b'(e_{T-2}) - c'(e_{T-2}) \\ &= [\beta b'(e_{T-1}) - c'(e_{T-1})] \left( -\frac{de_T}{de_{T-2}} \right) + \beta [b'(e_{T-1}) - c'(e_{T-1})] \left( -\frac{de_{T-1}}{de_{T-2}} \right) \\ &> \beta b'(e_{T-1}) - c'(e_{T-1}). \end{aligned}$$

To understand the last inequality note that by Lemma 4 we have  $\frac{de_T}{de_{T-2}} \leq 0$  and  $\frac{de_{T-1}}{de_{T-2}} < 0$ , and in addition  $\frac{de_{T-1}}{de_{T-2}} + \frac{de_T}{de_{T-2}} = -1$  (by Lemma 2). We conclude from both cases that either  $e_{T-2} < e_{T-1} \leq e_{max}(\beta)$  or we have a corner solution  $e_{T-2} = e_{T-1} = e_T = e_{max}(\beta)$ .

Continuing the iteration and using the first-order condition for self  $\tau < T$  for interior solutions,

$$\beta b'(e_\tau) - c'(e_\tau) = \sum_{t=\tau+1}^T \left[ \beta (b'(e_t) - c'(e_t)) \left( -\frac{de_t}{de_\tau} \right) \right], \quad (15)$$

while applying that  $\sum_{t=\tau+1}^T \frac{de_t}{de_\tau} = -1$ , gives  $e_1 \leq \dots \leq e_T \leq e_{max}(\beta)$ . Equality arises only if  $e_T = e_{max}(\beta)$ . In the latter case, a corner solution for the effort may arise for a number of periods leading up to  $T$ :  $e_{\underline{t}} = \dots = e_T = e_{max}(\beta)$ , where  $\underline{t} > 1$  because  $G < T e_{max}(\beta)$ . ■

### A.3.3 The self-0 preferred effort $e_0^*$ cannot be implemented in every period

Suppose that the individual faces (exogenously) the weekly goal  $G = T e_0^*$ . The first result covers both sophisticated and (partially) naïve individuals and formalizes the intuition for effort substitution given in the main text. The second result covers sophisticates only and shows that effort substitution leads to too little effort initially and too much effort in the end.

**Proposition 4** *Suppose  $e_0^* < e_{max}(\beta) \leq e_{max}(\hat{\beta})$ . Then, with daily goals, self 0 sets a goal  $g_t^* = e_0^*$  for each period and each goal is achieved. However, effort  $e_0^*$  is not implementable in each period with the weekly goal  $G = T e_0^*$ .*

**Proof.**

The first part follows from Proposition 1. We now show that the individual cannot implement  $e_0^*$  in each period with the weekly goal  $G = T e_0^*$ , even though  $G < T e_{max}(\beta)$ . Solving backward, we start with the behavior of self  $T$ . If all previous selves chose  $e_0^*$ , then the problem of self  $T$  would look exactly like the problem under daily goals. That is, self  $T$  would provide  $e_0^*$ . Now suppose that at least one previous self  $t < T$  worked less hard than  $e_0^*$ , so that  $\sum_{t=1}^{T-1} e_t + e_0^* < G = T e_0^*$ . Self  $T$  responds with  $e_T = \min\{G - \sum_{t=1}^{T-1} e_t, e_{max}(\beta)\}$  (Lemma 1).

We next show that some self  $t < T$  has an incentive to deviate from  $e_t = e_0^*$ , given the anticipated response of self  $T$  to such a deviation. Specifically, we show that self  $T - 1$  has an incentive to deviate if none of the previous selves already deviated. To take into account the possibility of naïveté, denote by  $\hat{e}_{T,t}$  the belief that a self  $t < T$  holds about the effort of self  $T$  based on  $\hat{\beta}$ . Note that  $e_{max}(\hat{\beta}) \geq e_{max}(\beta) > e_0^*$  implies that for a small deviation by self  $T - 1$  to  $e_{T-1} = e_0^* - \epsilon$  we have

$$(T - 2) e_0^* + e_{T-1} + e_{max}(\hat{\beta}) > T e_0^* = G, \quad (16)$$

and self  $T - 1$  anticipates the effort response by self  $T$  to be  $d\hat{e}_{T,T-1}/de_{T-1} = -1$ . That is, self  $T - 1$  believes that self  $T$  will fully make up for the  $\epsilon$  shortfall in effort. Note that for  $\epsilon$  bounded away from zero (16) may hold while at the same time  $(T - 2) e_0^* + e_{T-1} + e_{max}(\beta) < T e_0^*$ . In this case, the actual effort response  $de_T/de_{T-1} > -1$  is not sufficient to make up for the  $\epsilon$  shortfall in effort and the naïve individual will in the end not achieve the goal  $G$ .

Now consider the utility impact on self  $T - 1$  of a marginal deviation from  $e_0^*$ . The left-derivative of the utility of self  $T - 1$  at decision  $e_{T-1} = e_0^*$  is:

$$\beta b'(e_{T-1}) - c'(e_{T-1}) + \beta \underbrace{[b'(\hat{e}_{T,T-1}) - c'(\hat{e}_{T,T-1})]}_{=0 \text{ at } e_T = e_0^*} \frac{d\hat{e}_{T,T-1}}{de_{T-1}} < 0.$$

That is, deviating from  $e_0^*$  by lowering effort  $e_{T-1}$  increases the utility of self  $T - 1$ . Hence,  $e_0^*$  is not implementable in each period with a weekly goal. ■



**Lemma 6** *Suppose  $\hat{\beta} = \beta$  and  $G = T e_0^* < T e_{max}(\beta)$ . Self  $T$  (self 1) works more (less) than is optimal from the perspective of self 0:  $e_T > e_0^*$  and  $e_1 < e_0^*$ .*

**Proof.**

In the proof of Proposition 4 we showed that  $e_T > e_0^*$ . From Lemma 5, we know that  $e_1 < e_2 \leq \dots \leq e_T \leq e_{max}(\beta)$ . If all selves  $t > 1$  provide  $e_{max}(\beta)$ , then  $e_1 < e_0^*$  as  $G = T e_0^* < T e_{max}(\beta)$ . If only selves  $t \geq \underline{t}$  provide  $e_{max}(\beta)$ , the part of the goal that selves  $t < \underline{t}$  have to fulfill requires an average effort less than  $e_0^*$ . Together with  $e_1 < \dots < e_{\underline{t}-1} < e_{\underline{t}} = \dots = e_T = e_{max}(\beta)$  (from the proof of Lemma 5) this implies that  $e_1 < e_0^*$ . The final case is where  $e_T \in (e_0^*, e_{max}(\beta))$ . Because  $\sum_{t=1}^{T-1} e_t < (T-1)e_0^*$ , Lemma 5 implies that  $e_1$  is less than the average of  $e_t$ , which in turn is less than  $e_0^*$ . ■

## A.4 Proof of Proposition 2

We draw on the results in Appendix A.3. Lemma 2 gives that the goal is achieved. Part 1 is obtained by extending Lemma 5 (to show that effort is (weakly) increasing over time) and Lemma 6 (to show that  $e_T > e_0^*$ ). Part 2 follows from Lemma 3 and the utility of self 0 being increasing in effort  $e_t$  up to  $e_0^*$ .

### Step 1: Relation between $e_t$ and $G$

The following lemma provides some structure on how effort levels  $e_t$  respond to a change in the weekly goal  $G$ .

**Lemma 7**  *$\sum_{t=1}^T \frac{de_t}{dG} = 1$  and  $0 \leq \frac{de_t}{dG} \leq 1$ . For  $G < T e_{max}(\beta)$  we have  $\frac{de_1}{dG} > 0$ . If, in addition,  $\beta b'''(e) - c'''(e) \leq 0$  and  $b'''(e) - c'''(e) \leq 0$ , then there exists  $\underline{t} \in \{2, \dots, T+1\}$  such that  $e_{\underline{t}-1} < e_{max}(\beta)$  and  $\frac{de_{\underline{t}-1}}{dG} > \frac{de_{\underline{t}-2}}{dG} > \dots > \frac{de_1}{dG}$ .*

**Proof.**

Let  $v_t(e_t) \equiv \beta b(e_t) - c(e_t)$  and  $w_t(e_t) \equiv \beta [b(e_t) - c(e_t)]$ . Taking the total derivative of the first-order condition (15), and noting that  $\frac{d^2 e_t}{de_\tau dG} = 0$ , we get (dropping arguments of  $v(\cdot)$  and  $w(\cdot)$  to facilitate exposition)

$$\frac{de_{T-t}}{dG} = \sum_{\tau=T-(t-1)}^T w_\tau'' \left[ \frac{\left( -\frac{de_\tau}{de_{T-t}} \right) \frac{de_\tau}{dG}}{v_{T-t}''} \right]$$

If  $T$  is not a corner solution ( $e_T < e_{max}(\beta)$ ), recursively plugging in yields:

$$\begin{aligned} \frac{de_{T-1}}{dG} &= \frac{w''_T}{v''_{T-1}} \underbrace{\left(-\frac{de_T}{de_{T-1}}\right)}_{=1} \frac{de_T}{dG}, \\ \frac{de_{T-2}}{dG} &= \frac{w''_{T-1}}{v''_{T-2}} \left(-\frac{de_{T-1}}{de_{T-2}}\right) \frac{de_{T-1}}{dG} + \frac{w''_T}{v''_{T-2}} \left(-\frac{de_T}{de_{T-2}}\right) \frac{de_T}{dG}, \\ \frac{de_{T-3}}{dG} &= \frac{w''_{T-2}}{v''_{T-3}} \left(-\frac{de_{T-2}}{de_{T-3}}\right) \frac{de_{T-2}}{dG} + \frac{w''_{T-1}}{v''_{T-3}} \left(-\frac{de_{T-1}}{de_{T-3}}\right) \frac{de_{T-1}}{dG} + \frac{w''_T}{v''_{T-3}} \left(-\frac{de_T}{de_{T-3}}\right) \frac{de_T}{dG}, \\ \frac{de_{T-4}}{dG} &= \frac{w''_T}{v''_{T-4}} \frac{de_T}{dG} \left[ -\frac{de_T}{de_{T-4}} - \frac{w''_{T-1}}{v''_{T-1}} \frac{de_{T-1}}{de_{T-4}} - \frac{w''_{T-2}}{v''_{T-2}} \frac{de_{T-2}}{de_{T-4}} \left( -\frac{de_T}{de_{T-2}} - \frac{w''_{T-1}}{v''_{T-1}} \frac{de_{T-1}}{de_{T-2}} \right) \right. \\ &\quad \left. - \frac{w''_{T-3}}{v''_{T-3}} \frac{de_{T-3}}{de_{T-4}} \left( -\frac{de_T}{de_{T-3}} - \frac{w''_{T-1}}{v''_{T-1}} \frac{de_{T-1}}{de_{T-3}} - \frac{w''_{T-2}}{v''_{T-2}} \frac{de_{T-2}}{de_{T-3}} \left( -\frac{de_T}{de_{T-2}} - \frac{w''_{T-1}}{v''_{T-1}} \frac{de_{T-1}}{de_{T-2}} \right) \right) \right]. \end{aligned}$$

If  $T$  is a corner solution ( $e_T = e_{max}(\beta)$ ), start the above argument at the first interior solution  $e_{t-1} < e_{max}(\beta)$  (which exists because  $G < T e_{max}(\beta)$ ). Continuing in this way and noting that  $\frac{de_t}{de_\tau} \leq 0$  for  $\tau < t$  (Lemma 4) gives that for  $t < T$  (if  $e_T < e_{max}(\beta)$ ) or  $t < \underline{t} - 1$  (otherwise) either  $\frac{de_t}{dG} = 0$  or  $\text{sign}\left(\frac{de_t}{dG}\right) = \text{sign}\left(\frac{de_{t+1}}{dG}\right)$ . Further,  $\sum_{t=1}^T e_t = G$  implies that  $\sum_{t=1}^T \frac{de_t}{dG} = 1$ . Since all derivatives have the same sign, or are zero in case of a corner solution, and since they sum up to one, it follows that each derivative must lie between 0 and 1. From Lemma 5 it follows that for  $G < T e_{max}(\beta)$  we have an interior solution  $e_1 < e_{max}(\beta)$  and hence  $\frac{de_1}{dG} > 0$ .

We next show that for interior solutions  $e_T < e_{max}(\beta)$ , we have  $\frac{de_T}{dG} > \frac{de_{T-1}}{dG} > \dots > \frac{de_1}{dG}$ . Note that  $\frac{w''_t(e_t)}{v''_t(e_t)} < 1$ , because  $w'''(\cdot), v'''(\cdot) \leq 0$  ensures that  $v''(e_t) < w''(e_t) < 0$ . Moreover,  $\sum_{t=\tau+1}^T \left(-\frac{de_t}{de_\tau}\right) = 1$  and  $e_t > e_\tau$  for  $t > \tau$ . The result follows using the fact that  $-\frac{de_t}{de_\tau} > 0$  in the recursive definition of  $\frac{de_{T-t}}{dG}$  above. Taking account of possible corner solutions, we get that there exists  $\underline{t} \in \{2, \dots, T+1\}$  such that  $e_{\underline{t}-1} < e_{max}(\beta)$  (Lemma 5) and  $\frac{de_{\underline{t}-1}}{dG} > \frac{de_{\underline{t}-2}}{dG} > \dots > \frac{de_1}{dG}$ . ■

## Step 2: Interior solutions ( $e_t < e_{max}(\beta)$ )

Given that self 0 has to set a weekly goal, he chooses  $G$  to maximize his utility  $\beta \left( \sum_{t=1}^T b(e_t^b(G)) - c(e_t^b(G)) \right)$ , which then determines for all dates  $t = 1, \dots, T-1$  the effort  $e_t^b(G)$  through the system of first-order conditions (15). The effort  $e_T^b(G)$  is then pinned down by  $\sum_{t=1}^T e_t^b(G) = G$  (by Lemma 2, because  $G \leq T e_0^*$ ). The first-order condition for the optimal goal  $G^*$  is

given by:

$$\beta \sum_{t=1}^T \left[ (b'(e_t^b(G^*)) - c'(e_t^b(G^*))) \frac{de_t^b(G^*)}{dG} \right] = 0. \quad (17)$$

We now proceed to restate (17) by substituting in from the first-order conditions of selves  $t > 0$ . To facilitate exposition, we write  $e_t$  instead of  $e_t^b(G)$ . Rearranging (13) yields:

$$\beta [b'(e_{T-1}) - c'(e_{T-1})] = \beta [b'(e_T) - c'(e_T)] + (1 - \beta) c'(e_{T-1}). \quad (18)$$

Note that future effort  $e_T + e_{T-1}$  has to add up to the remaining goal  $G_{T-1} = G_{T-2} - e_{T-2}$ . Holding constant  $G_{T-2}$ , we thus have  $\frac{de_T}{de_{T-2}} + \frac{de_{T-1}}{de_{T-2}} = -1$  (by Lemma 2). Using this fact and substituting (18) into (14), gives

$$\begin{aligned} & \beta [b'(e_{T-2}) - c'(e_{T-2})] \\ &= \beta [b'(e_T) - c'(e_T)] + (1 - \beta) c'(e_{T-1}) \left( - \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \right) + (1 - \beta) c'(e_{T-2}). \end{aligned}$$

Similarly, for self  $T - 3$  we get:

$$\begin{aligned} & \beta [b'(e_{T-3}) - c'(e_{T-3})] \\ &= \beta [b'(e_T) - c'(e_T)] \cdot \overbrace{(-1)}^{=-1} \cdot \left( \frac{de_T}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} + \frac{de_{T-1}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} + \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right) \\ &+ (1 - \beta) c'(e_{T-1}) \left[ \left( - \frac{de_{T-1}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right) + \left( - \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \right) \left( - \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right) \right] \\ &+ (1 - \beta) c'(e_{T-2}) \left( - \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-4}=\text{const}} \right) + (1 - \beta) c'(e_{T-3}). \end{aligned} \quad (19)$$

Note that

$$\frac{de_{T-1}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \neq \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \cdot \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}}.$$

To see this, consider a given remaining goal  $G_{T-3}$ . The remaining goal for self  $T - 1$  then satisfies  $G_{T-1} = G_{T-3} - e_{T-2} - e_{T-3}$  and thus

$$\frac{de_{T-1}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} = \frac{de_{T-1}}{dG_{T-1}} \cdot \frac{dG_{T-1}}{de_{T-3}} = \left( - \frac{de_{T-1}}{dG_{T-1}} \right) \cdot \left( 1 + \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right).$$

Using  $G_{T-1} = G_{T-2} - e_{T-2}$ , we thus get

$$\frac{de_{T-1}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} = \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \cdot \left( 1 + \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right). \quad (20)$$

More generally, for integers  $m, n$ ,  $0 < m < n < T$ ,

$$\frac{de_{T-m}}{de_{T-n}} \Big|_{G_{T-n}=\text{const}} = \frac{de_{T-m}}{de_{T-m-1}} \Big|_{G_{T-m-1}=\text{const}} \cdot \left( 1 + \sum_{k=m+1}^{n-1} \frac{de_{T-k}}{de_{T-n}} \Big|_{G_{T-n}=\text{const}} \right). \quad (21)$$

Using (20) we can rewrite (19) as

$$\begin{aligned} \beta [b'(e_{T-3}) - c'(e_{T-3})] &= \beta [b'(e_T) - c'(e_T)] \\ &+ (1 - \beta) c'(e_{T-1}) \left( - \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \right) + (1 - \beta) c'(e_{T-2}) \left( - \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right) \\ &+ (1 - \beta) c'(e_{T-3}). \end{aligned}$$

Similarly, for self  $T - 4$  we get:

$$\begin{aligned} \beta [b'(e_{T-4}) - c'(e_{T-4})] &= \beta [b'(e_T) - c'(e_T)] \\ &+ (1 - \beta) c'(e_{T-1}) \left[ \left( - \frac{de_{T-1}}{de_{T-4}} \Big|_{G_{T-4}=\text{const}} \right) \right. \\ &\quad \left. + \left( - \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \right) \left( - \frac{de_{T-2}}{de_{T-4}} \Big|_{G_{T-4}=\text{const}} - \frac{de_{T-3}}{de_{T-4}} \Big|_{G_{T-4}=\text{const}} \right) \right] \\ &+ (1 - \beta) c'(e_{T-2}) \left[ \left( - \frac{de_{T-2}}{de_{T-4}} \Big|_{G_{T-4}=\text{const}} \right) + \left( - \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right) \left( - \frac{de_{T-3}}{de_{T-4}} \Big|_{G_{T-4}=\text{const}} \right) \right] \\ &+ (1 - \beta) c'(e_{T-3}) \left( - \frac{de_{T-3}}{de_{T-4}} \Big|_{G_{T-4}=\text{const}} \right) + (1 - \beta) c'(e_{T-4}). \end{aligned}$$

Using (21) we can rewrite this as

$$\begin{aligned} \beta [b'(e_{T-4}) - c'(e_{T-4})] &= \beta [b'(e_T) - c'(e_T)] + (1 - \beta) c'(e_{T-1}) \left( - \frac{de_{T-1}}{de_{T-2}} \Big|_{G_{T-2}=\text{const}} \right) \\ &+ (1 - \beta) c'(e_{T-2}) \left( - \frac{de_{T-2}}{de_{T-3}} \Big|_{G_{T-3}=\text{const}} \right) + (1 - \beta) c'(e_{T-3}) \left( - \frac{de_{T-3}}{de_{T-4}} \Big|_{G_{T-3}=\text{const}} \right) \\ &+ (1 - \beta) c'(e_{T-4}). \end{aligned}$$

Continuing in this way, we get for  $n = 1, \dots, T - 1$ ,

$$\begin{aligned} \beta [b'(e_{T-n}) - c'(e_{T-n})] &= \beta [b'(e_T) - c'(e_T)] + (1 - \beta) \sum_{m=1}^{n-1} \left\{ c'(e_{T-m}) \left[ - \frac{de_{T-m}}{de_{T-m-1}} \Big|_{G_{T-m-1}=\text{const}} \right. \right. \\ &\quad \left. \left. + \sum_{k=m+1}^{n-1} \left( - \frac{de_{T-m}}{de_{T-m-1}} \Big|_{G_{T-m-1}=\text{const}} \right) \left( \frac{de_{T-k}}{de_{T-n}} \Big|_{G_{T-n}=\text{const}} \right) \right] \right\} + (1 - \beta) c'(e_{T-n}). \end{aligned}$$

Using (21) we can rewrite this as

$$\begin{aligned} \beta [b'(e_{T-n}) - c'(e_{T-n})] &= \beta [b'(e_T) - c'(e_T)] \\ &+ (1 - \beta) \sum_{m=1}^{n-1} \left[ c'(e_{T-m}) \left( - \frac{d e_{T-m}}{d e_{T-m-1}} \Big|_{G_{T-m-1}=\text{const}} \right) \right] + (1 - \beta) c'(e_{T-n}). \end{aligned} \quad (22)$$

Plugging (22) into (17) and using  $\sum_{t=1}^T \frac{d e_t}{d G} = 1$ , yields:

$$\begin{aligned} 0 &= \beta [b'(e_T^b(G^*)) - c'(e_T^b(G^*))] + (1 - \beta) \sum_{n=1}^{T-1} \left[ c'(e_{T-n}^b(G^*)) \frac{d e_{T-n}^b(G^*)}{d G} \right. \\ &\quad \left. + c'(e_{T-n}^b(G^*)) \left( - \frac{d e_{T-n}^b(G^*)}{d e_{T-n-1}^b(G^*)} \Big|_{G_{T-n-1}=\text{const}} \right) \cdot \sum_{m=n+1}^{T-1} \left( \frac{d e_{T-m}^b(G^*)}{d G} \right) \right]. \end{aligned}$$

We can rewrite this as

$$\begin{aligned} 0 &= \beta [b'(e_T^b(G^*)) - c'(e_T^b(G^*))] + (1 - \beta) c'(e_1^b(G^*)) \frac{d e_1^b(G^*)}{d G} + \Omega, \quad \text{where} \quad (23) \\ \Omega &= (1 - \beta) \sum_{n=1}^{T-2} \left[ c'(e_{T-n}^b(G^*)) \frac{d e_{T-n}^b(G^*)}{d G} \right. \\ &\quad \left. + c'(e_{T-n}^b(G^*)) \left( - \frac{d e_{T-n}^b(G^*)}{d e_{T-n-1}^b(G^*)} \Big|_{G_{T-n-1}=\text{const}} \right) \cdot \sum_{m=n+1}^{T-1} \left( \frac{d e_{T-m}^b(G^*)}{d G} \right) \right]. \end{aligned}$$

First note that  $\Omega > 0$ : By assumption,  $e_0^* < e_{\max}(\beta)$ . By Lemma 7,  $\frac{d e_t^b(G^*)}{d G} \geq 0$ , with strict inequality for at least  $t = 1$ . Further,  $\frac{d e_{T-n}^b(G^*)}{d e_{T-n-1}^b(G^*)} \Big|_{G_{T-n-1}=\text{const}} \leq 0$ . Thus, (23) together with the fact that  $\beta [b'(e_0^*) - c'(e_0^*)] = 0$  and that  $b'(e) - c'(e)$  is strictly decreasing, implies that  $e_T > e_0^*$ .<sup>23</sup>

### Step 3: Corner solutions

**Lemma 8** *There may exist a corner solution for the effort in periods  $\underline{t}, \dots, T$ , where  $\underline{t} > 1$ .*

*The effort schedule then is  $(e_1^b, \dots, e_{\underline{t}-1}^b, e_{\max}(\beta), \dots, e_{\max}(\beta))$ , where for  $\underline{t} > 2$  we have  $e_1^b < e_2^b < \dots < e_{\underline{t}-2}^b < e_0^* < e_{\underline{t}-1}^b$  with  $e_{\tau}^b$ ,  $\tau < \underline{t}$ , characterized by*

$$\beta b(e_{\tau}^b) - c'(e_{\tau}^b) = \sum_{t=\tau+1}^{\underline{t}-1} \left[ \beta (b(e_t^b) - c'(e_t^b)) \left( - \frac{d e_t^b}{d e_{\tau}^b} \right) \right]. \quad (24)$$

For  $\underline{t} = 2$ ,  $e_1^b = e_0^*$ .

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<sup>23</sup>For time-consistent preferences, i.e.  $\beta = 1$ , we would have  $e_T = e_0^*$ .

**Proof.**

The result follows from Lemma 5 and the following argument. If selves  $\underline{t}, \dots, T$  provide  $e_{max}(\beta)$ , then from the perspective of self  $\underline{t} - 1$ , the effort of future selves is fixed. So self  $\underline{t} - 1$  either sticks to the effort according to the plan of self 0 or he makes-up for a previous shortfall in the same way as described for self  $T$  above. Hence, we can redo the arguments described above, substituting  $T$  and  $G$ , respectively, with the final period  $T' = \underline{t} - 1$  now being the last period with an interior solution, and the part of the goal to be achieved by  $T'$  being  $G - T'e_{max}(\beta)$ . ■

## A.5 Proof of Proposition 3

We show parts 1 and 2 in four steps.

### Step 1: Effort profile with interior solution

**Lemma 9** *Suppose  $\beta b'''(e) - c'''(e) \leq 0$  and  $b'''(e) - c'''(e) \leq 0$ . Then, for an interior solution with  $e_t^b(G^*) < e_{max}(\hat{\beta})$  for  $t = 1, \dots, T$  we have  $T e_0^* > \sum_{t=1}^T e_t^b(G^*)$ , where  $e_t^b(G^*)$  and  $G^*$  are characterized by (15) and (17).*

**Proof.**

Let  $q(e) \equiv b(e) - c(e)$ . Rewriting the first-order condition for an interior solution (17) using the fact that  $q'(e_0^*) \equiv b'(e_0^*) - c'(e_0^*) = 0$ , we get

$$\sum_{t=1}^T \left[ q'(e_t^b(G^*)) \frac{d e_t^b(G^*)}{d G} \right] = q'(e_0^*).$$

Rewriting, using  $\sum_{t=1}^T \frac{d e_t^b(G^*)}{d G} = 1$ :

$$\begin{aligned} q' \left( \sum_{t=1}^T e_0^* \frac{d e_t^b(G^*)}{d G} \right) &= \sum_{t=1}^T \left[ q'(e_t^b(G^*)) \frac{d e_t^b(G^*)}{d G} \right] \\ &\leq q' \left( \sum_{t=1}^T e_t^b(G^*) \frac{d e_t^b(G^*)}{d G} \right), \end{aligned}$$

where the last line follows from Jensen's inequality because  $q'(\cdot)$  is concave. Hence, from  $q''(\cdot) < 0$  it follows that

$$\sum_{t=1}^T (e_0^* - e_t^b(G^*)) \frac{d e_t^b(G^*)}{d G} \geq 0.$$

For an interior solution  $e_T$  then  $1 > \frac{de_T}{dG} > \frac{de_{T-1}}{dG} > \dots > \frac{de_1}{dG} > 0$  (by Lemma 7) gives that:

$$\frac{de_T^b(G^*)}{dG} \sum_{t=1}^T (e_0^* - e_t^b(G^*)) > \sum_{t=1}^T (e_0^* - e_t^b(G^*)) \frac{de_t^b(G^*)}{dG} \geq 0.$$

Hence, we conclude that

$$T e_0^* > \sum_{t=1}^T e_t^b(G^*).$$

■

### Step 2: Effort profiles with a corner solution

Lemma 8 characterizes effort profiles with a corner solution.

### Step 3: Implemented effort profile

#### Lemma 10

- (i) For  $\beta$  sufficiently close to 1, self 0 prefers to implement  $(e_1^b, \dots, e_t^b)$  (characterized in Lemma 9) rather than any other  $(e_1^{b'}, \dots, e_{\tau-1}^{b'}, e_{max}(\beta), \dots, e_{max}(\beta))$  (characterized in Lemma 8).
- (ii) For  $\beta$  sufficiently close to the cutoff  $\check{\beta} : e_{max}(\check{\beta}) = e_0^*$ , self 0 prefers to implement  $(e_0^*, e_{max}(\beta), \dots, e_{max}(\beta))$  rather than an interior solution  $(e_1^b, \dots, e_t^b)$  or a (partially) interior solution with  $\underline{t} \in \{3, \dots, T+1\}$  such that  $e_1 < e_2 < \dots < e_{\underline{t}-1} < e_{\underline{t}}$  and  $e_t = e_{max}(\beta)$  for  $t \in \{\underline{t}, \dots, T\}$ .

#### Proof.

(i) From (15) and (17) it follows that  $e_t^b(G^*(1)) = e_0^*$ , while for  $\beta = 1$  we have  $e_{max} > e_0^*$ . Hence,  $U_0(e_1^b(G^*(1)), \dots, e_T^b(G^*(1))) = U_0(e_0^*, \dots, e_0^*) > U_0(e_1^{b'}, \dots, e_{\tau}^{b'}, e_{max}, \dots, e_{max}) = U_0(e_0^*, \dots, e_0^*, e_{max}, \dots, e_{max})$ . The result follows from the intermediate value theorem because the utility function is continuous in all arguments and  $e_{max}$  is a continuous, increasing function of  $\beta$ .

(ii)  $U_0(e_0^*, e_{max}(\beta), \dots, e_{max}(\beta))$  is continuous in  $\beta$  and has limit  $U_0(e_0^*, e_0^*, \dots, e_0^*)$  as  $\beta \rightarrow \check{\beta}$ , where  $e_{max}(\check{\beta}) = e_0^*$ . The alternative of a (partially) interior solution may not be feasible

because it calls for  $e_T > e_{max}(\beta)$  ( $e_{t-1} > e_{max}(\beta)$ ). But even if it is feasible, the utility does not converge to  $U_0(e_0^*, e_0^*, \dots, e_0^*)$  as  $\beta \rightarrow \check{\beta}$ . For any  $\beta \in (\check{\beta})$  one cannot implement  $e_0^*$  in every period, as shown in Proposition 4, and from the first-order condition (13) for an interior solution  $e_{T-1}$  it follows that for  $\beta$  bounded away from 1 we will have  $e_{T-1}$  bounded away from  $e_T$  (for a partially interior solution, the same argument applies to the periods with interior solutions). ■

## A.6 A (partially) naïve individual facing a weekly goal

**Proposition 5** *Suppose a (partially) naïve individual ( $1 \geq \hat{\beta} > \beta$ ) sets a weekly goal  $G$ .*

1. For  $\hat{\beta} = 1$  (full naïveté),  $G^* = T e_0^*$  and

(a) all selves expect a constant effort pattern: self  $t = 0$  expects  $\hat{e}_{\tau,0} = e_0^*$  and self  $t \in \{1, \dots, T-1\}$  expects  $\hat{e}_{\tau,t} = \min\{(T e_0^* - \sum_{k=1}^t e_k)/(T-t), e_{max}(1)\}$  for  $\tau = t+1, \dots, T$ ,

(b) actual effort falls short of the one-period ahead expectation:  $e_t < \hat{e}_{t,t-1}$ ,

(c) expectations about future effort increase relative to those held in the previous period:  $\hat{e}_{t+\tau,t} \geq \hat{e}_{t+\tau,t-1}$ , with strict inequality if  $\hat{e}_{t+\tau,t-1} < e_{max}(1)$ .

2. For  $\hat{\beta} < 1$  and  $e_{max}(\hat{\beta}) > e_0^*$ ,

(a) self 0 expects a (weakly) increasing effort pattern  $\hat{e}_{1,0} < \hat{e}_{2,0} \leq \dots \leq \hat{e}_{T,0} \leq e_{max}(\hat{\beta})$ , where  $\hat{e}_{1,0} < e_0^*$  and  $\hat{e}_{T,0} > e_0^*$ , except in the corner solution with  $\hat{e}_{1,0} = e_0^*$  and  $\hat{e}_{t,0} = e_{max}(\hat{\beta})$  for  $t > 1$ ,

(b) actual effort falls short of the one-period ahead expectation:  $e_t < \hat{e}_{t,t-1}$  for  $t = 2, \dots, T-1$ ,  $e_1 < \hat{e}_{1,0}$  (except in the corner solution with  $\hat{e}_{1,0} = e_0^*$  as here  $e_1 = \min\{e_0^*, e_{max}(\beta)\}$ ) and  $e_T \leq \hat{e}_{T,T-1}$ , with strict inequality if  $\hat{e}_{T,T-1} > e_{max}(\beta)$ ,

(c) self  $t \in \{1, \dots, T-1\}$  expects a (weakly) increasing effort pattern,

(d) expectations about future effort increase relative to those held in the previous period:  $\hat{e}_{t+\tau,t} \geq \hat{e}_{t+\tau,t-1}$ , with strict inequality if  $\hat{e}_{t+\tau,t-1} < e_{max}(\hat{\beta})$ .

3. For  $\hat{\beta} < 1$  and  $e_{max}(\hat{\beta}) \leq e_0^*$ ,  $G^* = T e_{max}(\hat{\beta})$ ,  $e_t = e_{max}(\beta) < \hat{e}_{t,\tau} = e_{max}(\hat{\beta})$ , for  $\tau < t$ .



**Proof.**

**1 Full naïveté** ( $\hat{\beta} = 1 > \beta$ ).  $G^* = T e_0^*$  and 1(a) are straightforward. 1(b) will be shown together with 2(b). 1(c) follows from 1(a) and 1(b).

**2 Partial naïveté** ( $1 > \hat{\beta} > \beta$ ). 2(a) follows from Proposition 2, because the case of a sophisticated individual ( $\hat{\beta} = \beta$ ) pins down the beliefs of self 0 about future effort.

**2(b) - 2(e) Actual effort and evolution of beliefs about effort.** The actual effort in periods  $\tau \in \{1, \dots, T\}$  satisfies

$$\beta b'(e_\tau) - c'(e_\tau) = \beta \sum_{t=\tau+1}^T \left[ (b'(\hat{e}_{t,\tau}) - c'(\hat{e}_{t,\tau})) \left( -\frac{d\hat{e}_{t,\tau}}{de_\tau} \right) \right], \quad (25)$$

where  $\hat{e}_{t,\tau}$  denotes the belief of self  $\tau$  about the effort of self  $t > \tau$ . Specifically, self  $\tau$  believes that a future self  $\tau' > \tau$  will choose  $\hat{e}_{\tau',\tau}$  to satisfy

$$\hat{\beta} b'(\hat{e}_{\tau',\tau}) - c'(\hat{e}_{\tau',\tau}) = \hat{\beta} \sum_{t=\tau'+1}^T \left[ (b'(\hat{e}_{t,\tau}) - c'(\hat{e}_{t,\tau})) \left( -\frac{d\hat{e}_{t,\tau}}{d\hat{e}_{\tau',\tau}} \right) \right]. \quad (26)$$

As long as self  $\tau$  believes that the goal will be reached (with future selves providing effort up to  $e_{max}(\hat{\beta})$ ), Lemmas 4 and 5 apply to the beliefs about future effort. If a self  $\tau$  concludes, after observing previous effort levels, that the goal will not be achieved even if he provides  $e_{max}(\beta)$  and all future selves provide  $e_{max}(\hat{\beta})$ , then  $e_\tau = e_t = e_{max}(\beta)$  for  $t = \tau + 1, \dots, T$  (Lemma 1).

- Consider  $\hat{e}_{1,0} \leq e_{max}(\beta)$ .

We first show that  $e_1 < \hat{e}_{1,0}$ . Denote by  $G_2 = G - e_1$  the part of the goal that selves  $t = 2, \dots, T$  need to achieve. Note that the beliefs of self 1 and self 0 about future effort  $e_t$  for  $t > 2$  coincide for a given  $G_2$ , because both believe that future selves will reason with  $\hat{\beta}$ . In particular, both agree on how future effort would change in response to a deviation from the solution  $e_1 = \hat{e}_{1,0}$  given by the implicit function (25) at  $\tilde{\beta} = \hat{\beta}$  and  $\tau = 1$ :

$$\tilde{\beta} b'(e_1) - c'(e_1) = \tilde{\beta} \sum_{t=2}^T \left[ (b'(\hat{e}_{t,0}) - c'(\hat{e}_{t,0})) \left( -\frac{d\hat{e}_{t,0}}{de_1} \right) \right]. \quad (27)$$

The left-hand side of (27) is strictly negative (i.e.,  $\hat{e}_{1,0} > e_1^*$ ). By way of contradiction, suppose it was non-negative. Note that for given  $G$ , a change in  $e_1$  implies a corresponding change in  $G_2 = G - e_1$ . Hence,  $-\frac{d\hat{e}_{t,0}}{de_1} = \frac{d\hat{e}_{t,0}}{dG_2}$ . Thus, the right-hand side of (27) gives the slope of the utility from a remaining goal  $G_2$  for periods  $t > 1$  from self 0 perspective. Now suppose the left-hand side of (27) was non-negative (i.e.,  $\hat{e}_{1,0} \leq e_1^*$ ). This would contradict optimal goal setting by self 0, because increasing both  $e_1$  and  $G_2$  (i.e. increasing  $G$ ) would increase the utility of self 0 and there is no conflict of interest with self 1. First, the utility of self 0 and 1 is increasing in  $e_1$  (with a zero slope for self 1 at  $\hat{e}_{1,0} = e_1^*$ ). Second, both selves agree about the utility from future effort levels  $e_2, \dots, e_T$ , and hence agree that raising  $G_2$  strictly increases utility as long as the right-hand side is positive.

By our assumptions on  $b(\cdot)$  and  $c(\cdot)$ , the left-hand side of (27) is strictly decreasing in  $e_1$ . The right-hand side of (27) must be strictly increasing in  $e_1$  because otherwise (27) could be satisfied at a lower  $e_1$ , which would increase the utility of self 1 (since  $\hat{e}_{1,0} > e_1^*$ ) and lead to a contradiction. Plugging  $\tilde{\beta} = \hat{\beta}$  into (27) pins down  $\hat{e}_{1,0}$ . Actual effort is determined by plugging in  $\tilde{\beta} = \beta$ . Starting from  $\hat{e}_{1,0}$ , since  $\beta < \hat{\beta}$ , the left-hand side of (27) is strictly less than the right-hand side and  $e_1$  needs to be reduced to satisfy (27). Thus,  $e_1 < \hat{e}_{1,0}$ .

This in turn gives  $G_2 = G - e_1 > G - \hat{e}_{1,0}$ . Applying Lemma 7 to  $T_2 = T - 1$  and  $G_2 = G - e_1$  implies that beliefs about future effort increase relative to those held by self 0.

- Consider  $\hat{e}_{1,0} > e_{max}(\beta)$ . Now 2(b) follows because the actual effort of self 1 is  $\min\{e_{max}(\beta), e_1\}$ , where  $e_1$  is the solution from the previous case. 2(c) follows by the same argument as before.

We now show that  $e_2 < \hat{e}_{2,1}$ . This is trivially the case if  $e_{max}(\beta) < \hat{e}_{2,1}$ . Consider  $e_{max}(\beta) \geq \hat{e}_{2,1}$  and denote by  $G_3 = G - e_1 - e_2$  the part of the goal that selves  $t = 3, \dots, T$  need to achieve. Note that the beliefs of self 2 and self 1 about future effort  $e_t$  for  $t > 3$  coincide for a given  $G_3$  and both agree on how future effort would change in response to a deviation

from the solution  $e_2 = \hat{e}_{2,1}$  for  $\tilde{\beta} = \hat{\beta}$  of the implicit function

$$\tilde{\beta} b'(e_2) - c'(e_2) = \tilde{\beta} \sum_{t=3}^T \left[ (b'(\hat{e}_{t,1}) - c'(\hat{e}_{t,1})) \left( -\frac{d\hat{e}_{t,1}}{de_2} \right) \right]. \quad (28)$$

Above we showed that  $\hat{e}_{2,1} > \hat{e}_{2,0} > \hat{e}_{1,0}$  and  $\hat{\beta} b'(\hat{e}_{1,0}) - c'(\hat{e}_{1,0}) < 0$ . Hence, the left-hand side of (28) is strictly negative. By our assumptions on  $b(\cdot)$  and  $c(\cdot)$ , it is strictly decreasing in  $e_2$ . Thus, the right-hand side of (28) must be strictly increasing in  $e_2$  because otherwise (28) could be satisfied at a lower  $e_2$ , which would increase the utility of self 2 (since  $\hat{e}_{2,1} > e_2^*$ ) and lead to a contradiction. Plugging  $\tilde{\beta} = \beta$  into (28), the left-hand side of (28) is strictly less than the right-hand side and  $e_2$  needs to be reduced to satisfy (28). Thus,  $e_2 < \hat{e}_{2,1}$ . By our above argument, beliefs about future effort increase. Repeating the arguments for  $t = 3, \dots, T - 1$  gives the result.

**3** ( $e_{max}(\hat{\beta}) \leq e_0^*$ ). Given belief  $\hat{\beta} > \beta$ , self 0 anticipates that future selves will provide at most  $e_{max}(\hat{\beta})$  (applying Lemma 3) and hence sets  $G^* = T e_{max}(\hat{\beta})$ . The actual effort pattern follows from the proof of Lemma 1: Effort in a given period can at most be  $e_{max}(\beta)$ . Self 1 believes that all future selves will provide at most  $e_{max}(\hat{\beta})$ . He provides up to  $e_{max}(\beta)$  to reduce the overall expected short-fall in effort relative to  $G$ . Hence  $e_1 = e_{max}(\beta) < e_{max}(\hat{\beta})$ . The same argument applies to all selves  $t > 1$ . ■

## B Parametric example

To illustrate our characterization of beliefs and effort, consider the following parametric example with  $b(e) = e$  and  $c(e) = e^2/2$ . Here,  $e_0^* = 1$ ,  $e_1^*(\beta) = \beta$ , and  $e_{max}^*(\beta) = 2\beta$ .

### B.1 Sophisticated individual

Effort is determined by recursively substituting in, starting with  $e_5$  and then setting as the anticipated effort for that period  $\hat{e}_{5,t} = e_5$  for  $t < 5$ , etc. Denote by  $G_t$  the remaining goal

at the start of period  $t$  (i.e.,  $G_1 = G$ ). Further, denote

$$\begin{aligned} \mathbf{a} &= 1 + \beta, & \mathbf{b} &= \beta(1 + \beta^2), & \mathbf{A} &= \mathbf{a}^2(1 + \beta^2) + \mathbf{b}^2, \\ \mathbf{B} &= \mathbf{a}^2 + \mathbf{b}, & \mathbf{C} &= \mathbf{B}^2 + \beta \mathbf{A}, & \text{and } \mathbf{D} &= \beta^2 \mathbf{B}^2 + \mathbf{A} \mathbf{B}^2. \end{aligned}$$

In this way, we obtain  $e_5 = \min\{G_5 - \sum_{t=1}^4 e_t, 2\beta\}$ ,  $e_4 = \begin{cases} \min\{\frac{\beta}{\mathbf{a}} G_4, 2\beta\} & \text{if } G_4 - e_4 < 2\beta, \\ \min\{G_4 - 2\beta, 2\beta\} & \text{if } \hat{e}_{5,4} = 2\beta, \end{cases}$

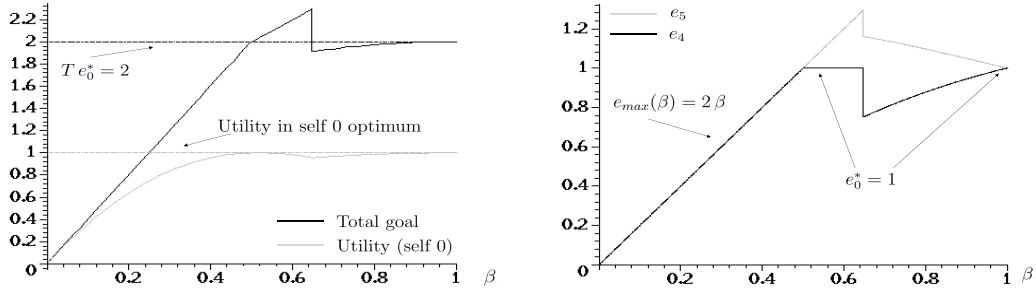
$$\begin{aligned} e_3 &= \begin{cases} \min\{\frac{\mathbf{b}}{\mathbf{B}} G_3, 2\beta\} & \text{if } G_3 - \hat{e}_{4,3} - e_3 < 2\beta \text{ and } \hat{e}_{4,3} < 2\beta, \\ \min\{\frac{\beta}{\mathbf{a}}(G_3 - 2\beta), 2\beta\} & \text{if } G_3 - 2\beta - e_3 < 2\beta \text{ and } \hat{e}_{5,3} = 2\beta, \\ \min\{G_3 - 4\beta, 2\beta\} & \text{if } \hat{e}_{4,3} = \hat{e}_{5,3} = 2\beta, \end{cases} \\ e_2 &= \begin{cases} \min\{\frac{\beta \mathbf{A}}{\mathbf{C}} G_2, 2\beta\} & \text{if } G_2 - \sum_{\tau=3}^5 \hat{e}_{\tau,2} - e_2 < 2\beta \text{ and } \hat{e}_{\tau,3} < 2\beta, \tau > 3, \\ \min\{\frac{\mathbf{b}}{\mathbf{B}}(G_2 - 2\beta), 2\beta\} & \text{if } G_2 - 2\beta - \sum_{\tau=3}^4 \hat{e}_{\tau,2} - e_2 < 2\beta, \hat{e}_{5,2} = 2\beta, \text{ and } \hat{e}_{4,2} < 2\beta, \\ \min\{\frac{\beta}{\mathbf{a}}(G_2 - 4\beta), 2\beta\} & \text{if } G_2 - 4\beta - e_2 < 2\beta \text{ and } \hat{e}_{4,2} = \hat{e}_{5,2} = 2\beta, \\ \min\{G_2 - 6\beta, 2\beta\} & \text{if } \hat{e}_{\tau,2} = 2\beta, \tau = 3, 4, 5, \end{cases} \\ e_1 &= \begin{cases} \min\left\{\frac{\beta \mathbf{D}}{\beta \mathbf{D} + \mathbf{C}^2} G, 2\beta\right\} & \text{if } G - \sum_{\tau=2}^5 \hat{e}_{\tau,1} - e_1 < 2\beta \text{ and } \hat{e}_{\tau,1} < 2\beta, \tau > 2, \\ \min\left\{\frac{\beta \mathbf{A}}{\mathbf{C}}(G - 2\beta), 2\beta\right\} & \text{if } G - 2\beta - \sum_{\tau=2}^4 \hat{e}_{\tau,1} - e_1 < 2\beta, \hat{e}_{5,1} = 2\beta, \text{ and} \\ & \hat{e}_{\tau,1} < 2\beta, \tau = 3, 4, \\ \min\left\{\frac{\mathbf{b}}{\mathbf{B}}(G - 4\beta), 2\beta\right\} & \text{if } G - 4\beta - \sum_{\tau=2}^3 \hat{e}_{\tau,1} - e_1 < 2\beta, \hat{e}_{4,1} = e_{5,1} = 2\beta, \text{ and} \\ & \hat{e}_{3,1} < 2\beta, \\ \min\left\{\frac{\beta}{\mathbf{a}}(G - 6\beta), 2\beta\right\} & \text{if } G - 6\beta - \hat{e}_{\tau,2} - e_1 < 2\beta, \text{ and } \hat{e}_{\tau,1} = 2\beta, \tau > 2 \\ \min\{G - 8\beta, 2\beta\} & \text{if } \hat{e}_{\tau,1} = 2\beta, \tau > 1. \end{cases} \end{aligned}$$

**Goal chosen by self 0.** Self 0 chooses  $G$  to solve

$$\max_G G - \sum_{t=1}^5 \hat{e}_{t,0}^2/2,$$

where for the case of a sophisticated individual  $\hat{e}_{t,0} = e_t$  as defined above. To find the solution, we consider the different subcases: The optimal goal among the goals that lead to (i) interior  $e_t < e_{\max}(\beta)$  for all  $t = 1, \dots, 5$ , (ii)  $e_5 = e_{\max}(\beta) > e_4 > \dots > e_1$ , (iii)  $e_5 = e_4 = e_{\max}(\beta) > e_3 > \dots > e_1$ , (iv)  $e_5 = e_4 = e_3 = e_{\max}(\beta) > e_2 > e_1$ , (v)  $e_5 = \dots = e_2 = e_{\max}(\beta) > e_1 = e_0^*$ , and (vi)  $e_t = e_{\max}(\beta)$  for all  $t = 1, \dots, 5$ . We then take the upper envelope of the utility functions for the different cases at their respective optimal goals to

Figure A1: Total goal, utility of self 0, and effort ( $T=2$ ).



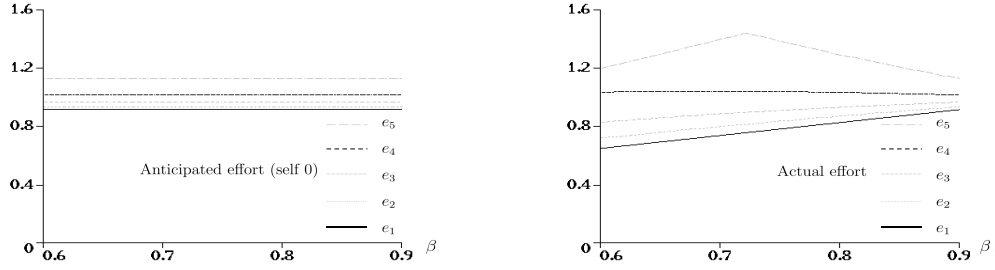
obtain the utility of self 0 as a function of  $\beta$ . Note that for a given  $\beta$  not all cases may be relevant. For example, for  $\beta < 0.5$ , case (v) is not feasible because  $e_0^* = 1 < e_{max}(\beta) = 2\beta$ . Further, note that the utility (and as illustrated below, the optimal goal and the associated effort profiles) are not monotone in  $\beta$ . The intuition is simple. For  $\beta = 0.5$ , self 0 can exactly implement  $e_t = e_0^*$  for all  $t = 1, \dots, 5$ , because  $e_0^* = e_{max}(\beta)$ . But for  $\beta = 0.5 + \epsilon$ , the problem of effort substitution arises and the individual can no longer implement a constant effort profile  $e_t = e_0^*$ . Setting  $e_t = e_{max}(\beta) = e_0^* + 2\epsilon$  for  $t > 1$  allows to commit to  $e_1 = e_0^*$ . For small  $\epsilon$ , utility is close to the self 0 optimum. In contrast, an interior solution (if at all feasible) results in a discrete downward shift in total effort and utility. For sufficiently high  $\beta$ , part 1 of Proposition 3 applies and  $G^* < T e_0^* = 5$ . Overall, utility is increasing up to  $\beta = 0.5$ , then drops and reaches its maximum again at  $\beta = 1$ .

For a simple illustration, consider the case with  $T = 2$  (just periods 4 and 5 remaining) shown in Figure A1. With an interior solution, the candidate goal for these two periods is  $G = \frac{(1+\beta)^2}{1+\beta^2}$ . For this to be feasible,  $e_5 = \frac{G}{1+\beta} \leq e_{max}(\beta) = 2\beta$ , i.e.,  $\beta$  has to exceed (approximately) 0.59. The utility from the corner solution  $e_5 = e_{max}(\beta)$  and  $e_4 = e_0^*$  is decreasing in  $\beta$  but dominates the interior solution for  $\beta$  less than (approximately) 0.65.

## B.2 Partially naïve individual

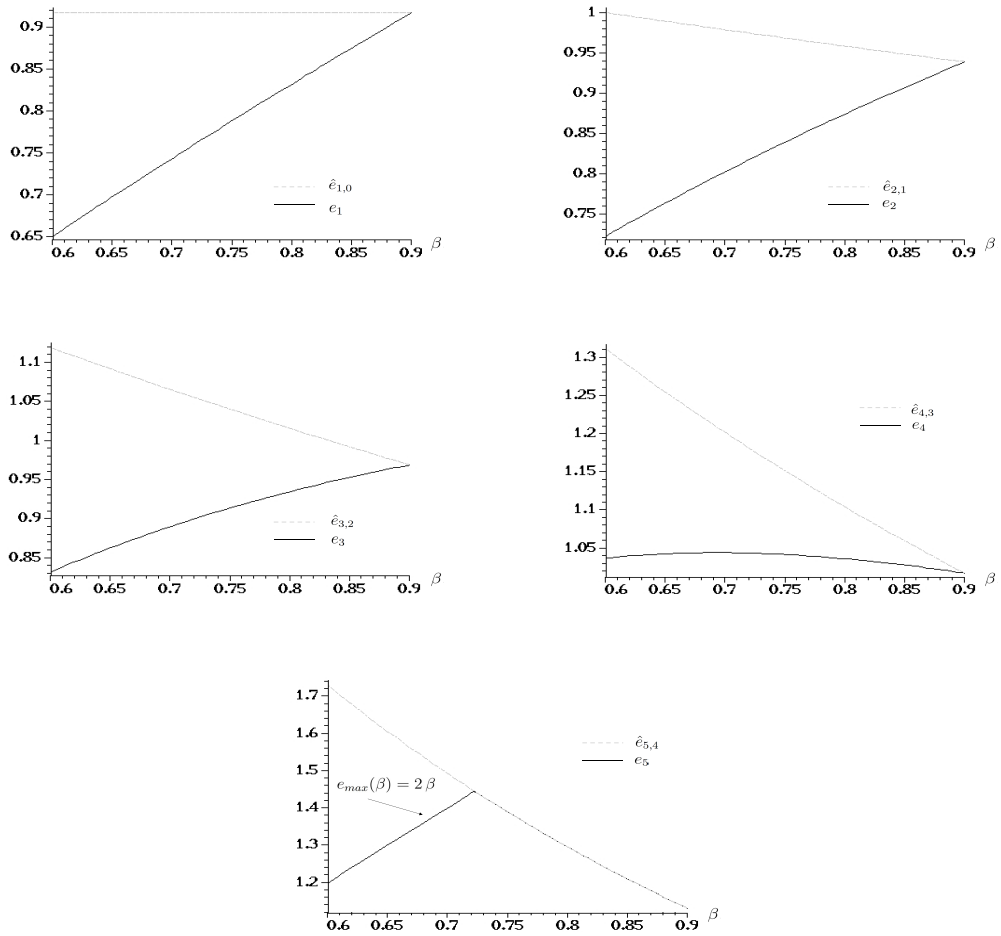
Just as in the case of a sophisticated individual, effort is determined by recursively substituting in. The difference is that self  $t$  applies the actual  $\beta$  to current period calculations but applies  $\hat{\beta}$  when determining anticipated effort  $\hat{e}_{\tau,t}$  for  $\tau > t$ . A partially naïve individual starts with the optimal goal for the belief  $\hat{\beta}$ ,  $G^*(\hat{\beta})$ . Obviously, the goal can only be achieved

Figure A2: Anticipated effort by self 0 and actual effort ( $\hat{\beta} = 0.9$ ).



if self 0 expects an interior effort in all periods  $\hat{e}_{5,0} > \hat{e}_{4,0} > \dots > \hat{e}_{1,0}$ . In our example, this happens for  $\hat{\beta} > 0.69$ . Otherwise, the goal will not be achieved because the individual overestimates the maximal implementable effort, and effort in the final period(s) will not be sufficient to reach  $G$ . In the main text, we show Figure 2 where the right panel fixes  $\hat{\beta} = 0.9$ . Self 0 sets goal  $G^*(\hat{\beta}) = 4.971$  and expects effort profile  $\hat{e}_{1,0} = 0.917$ ,  $\hat{e}_{2,0} = 0.939$ ,  $\hat{e}_{3,0} = 0.969$ ,  $\hat{e}_{4,0} = 1.017$ , and  $\hat{e}_{5,0} = 1.130$ . Figure A2 shows the effort that self 0 anticipates and the actual effort provided by selves  $t = 1, \dots, 5$ . Figure A3 shows that actual effort falls short of the one-period ahead expectations. In case of goal-nonachievement, the individual only realizes in the final period that the goal cannot be achieved because the anticipated effort  $\hat{e}_{t+\tau,t} < e_{max}(\hat{\beta}) = 1.8$  for  $t < 5$ .

Figure A3: Anticipated effort and actual effort ( $\hat{\beta} = 0.9$ ).



## C Relaxing the assumption of a constant cost function

The downward sloping pattern of daily goals in *Daily* (Figure 5) suggests that costs of effort/opportunity costs are increasing over the course of the week. Assuming that marginal costs increase in  $t$  implies that the self-0 preferred effort level for date  $t$ ,  $e_{t,0}^*$ , and the maximal implementable effort  $e_{max,t}(\beta)$  are decreasing in  $t$ . Thus, if there exists a cutoff period  $\tilde{t} > 1$  such that  $e_{max,1}(\hat{\beta}) \geq e_0^* > e_{max,t}(\hat{\beta})$  for  $t = \tilde{t}, \dots, T$ , the model predicts no difference in effort between *Daily* and *Weekly* from period  $\tilde{t}$  onwards. In both cases,  $e_t = e_{max}(\beta)$  for  $t \geq \tilde{t}$ . If  $\tilde{t} = 2$ ,  $e_1 = e_0^*$  in both treatments. If  $\tilde{t} > 2$ , we can apply the backward induction approach of the main model to determine whether an interior solution with anticipated effort levels  $\hat{e}_{t,0} \leq e_{max,t}(\hat{\beta})$  exists. Overall, compared to the main model,

- (i) we still predict that effort on Monday is likely to be lower in *Weekly* than in *Daily*, because effort substitution kicks in except in the case of a corner solution with anticipated effort profile  $(e_{1,0}^*, e_{max,2}(\hat{\beta}), \dots, e_{max,T}(\hat{\beta}))$ ;
- (ii) we predict less strongly that effort on Friday is higher in *Weekly* than in *Daily*, because a corner solution with  $e_T = e_{max,T}(\beta)$  is more likely than with constant effort costs.

## D Relaxing the assumption that all goals are evaluated in the last period

If daily goals already are evaluated at the end of each day but a weekly goal is only evaluated at the end of the week (period  $T + 1$ ), this creates an additional negative incentive effect under a weekly goal. This follows from comparing the maximal implementable effort (we omit the argument  $\beta$ , which is held fixed). For daily (narrow) goals,  $e_{max}^N$  in equation (4) is then defined by  $\beta b'(e_{max}^N) + 1 = c'(e_{max}^N)$ . For a weekly (broad) goal,  $e_{max}^B$  is still defined by  $\beta [b'(e_{max}^B) + 1] = c'(e_{max}^B)$ . Hence,  $e_{max}^B < e_{max}^N$ . As long as  $e_0^* \leq e_{max}^B < e_{max}^N$ , total effort  $G = T e_0^*$  is implementable under both goal setting formats. If  $e_{max}^B < e_0^* < e_{max}^N$ , then total effort  $G = T e_0^*$  would not be implementable under a weekly goal, but only  $T e_{max}^B$ .



## E Subjects

A total of 468 students from Aarhus University, Denmark, participated in our study. We recruited subjects for *Daily*, *Weekly*, *DailyRequirement*, and *WeeklyRequirement* in the Fall of 2013 through a large online study among first-year students at the faculty of Business and Social Sciences.<sup>24</sup> Subjects in *Daily(R)*, *Aggregated*, and *NoManipulation* were recruited in the Fall of 2014 through the subject pool of the experimental lab at Business and Social Sciences. About half of them were students at the faculty of Business and Social Sciences. We compare treatment pairs that used the same subject pool (Table A2).

As subjects in *Daily*, *Weekly*, *DailyRequirement*, and *WeeklyRequirement* were recruited from a previous online study, the 3 minute counting task took place few weeks before. These subjects however did receive a reminder about the task.

## F Additional control variables

Subjects for treatments *Daily*, *Weekly*, *WeeklyRequirement*, and *DailyRequirement* were recruited from a larger online survey experiment, described in Epper et. al (2018). From that previous study, we have additional control variables: *Self Control*: The 13-item Brief Self-Control scale (Tangney et al., 2004) *Grit*: 8-item Grit scale (Duckworth et al., 2007). *NarrowGoal*: A vignette question that reveals whether or not subjects would set narrow goals in a hypothetical exam preparation scenario. *Overconfidence*: Whether subjects overestimate their performance on a real-effort task relative to the performance of others. *High school grade*: Self-reported average grade in math and Danish in high school leaving exam. *Cognitive Reflection*: Cognitive reflection test (Frederick, 2005). *Mental Budget*: A 5-point Likert scale whether subjects divide their monthly budget into several separate budgets (such as budgets for housing, clothes, leisure expenditures, study related expenditures). *Loss Aversion*: Estimate of loss aversion parameter based on incentivized lottery task. *S-Shaped*: Estimate of shape of Prospect Theory value function.  $\tau$ : scaling parameter for error

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<sup>24</sup>The online study contained simple choice experiments and survey questions (Epper et al., 2018) and took place several weeks before the experiment. So tiredness did not affect the experiment described here. Yet, subjects were exposed to the counting task and thus had some experience (which did not affect the goal level or goal achievement; cf. Section 6).

standard deviation in risk preference estimates.

In addition, we do a principal component analysis on the 22 questions on goals and self-regulation from the previous study. They included questions about the type of goals students set for themselves, such as goals for course grades, or deadlines, questions about the goal setting process and potential mechanisms that help people stick to their goals, and questions regarding a subject's opinion about external, study-related commitment devices such as mandatory hand-in requirements or bets on study success. Following Jolliffe (2002, p. 133), we apply the Kaiser criterion of retaining only those principal components with variances greater than one and check that this procedure does not conflict with other selection criteria; namely that i) each component accounts for a sizeable part of the total variance (at least 5 percent), ii) cumulatively, the components account for at least 60 percent of the total variance, iii) the eigenvalues above and below the cut-off component are not too close. This procedure suggests 7 - 8 components. We retained 7 and checked that results are robust to using 8 components.

## G Task outsourcing

Conducting the study online has several advantages, but brings the possible disadvantage that we cannot control whether a subject outsourced the work after the goal setting stage. We believe this is unlikely because of the organizational hassle for a subject to find a low wage substitute. This is backed up by an analysis of the IP address, browser, operating system, and screen resolution at the stage when subjects set goals and when they started counting. We can exclude outsourcing to low-wage MTurkers because only two IP addresses are from outside of Europe.<sup>25</sup> Further, only for 16 subjects (3.6 percent of all) did we observe that they used a different computer in two different locations, where no location is the university.<sup>26</sup> This could have been due to subjects logging in, e.g. from their parents.

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<sup>25</sup>One subject contacted us about participating while abroad. The other participant used the same IP address for the goal setting and work stages. Only during the goal setting stage do subjects learn what the study is about.

<sup>26</sup>A change in the IP address with the same computer is not uncommon, because IP addresses are set dynamically and some students worked both at home and at the university.

But we cannot exclude that they employed somebody else to do the task. In any case, we would expect that those who outsourced their work would have exhibited different behavior (e.g. counted more tables). To check for this, we assume (very conservatively) that any change in IP address was due to outsourcing. However, a dummy that indicates a changing IP address reveals no significant relation to total effort (tobit marginal effect on censored outcome -12.11,  $p = 0.68$ ), goals set (-41.94,  $p = 0.16$ ), or number of logins (-0.46,  $p = 0.32$ ).

## H Checks for successful randomization and framing

By design, the total goal should not differ across *Daily(R)* and *Aggregated*, because subjects were randomized into different goal feedback frames only after having set their daily goals. Indeed, we observe no significant goal difference (permutation test,  $p = 0.914$ ; Table A16). In addition, the framing of goal feedback appears to have been successful. After having been given feedback about their weekly goal in *Aggregated*, subjects should have had that weekly goal in mind when making their effort choice and not the initial exercise of setting daily goals. That is, daily goals should be significant predictors of daily effort with the feedback format daily goals (*Daily(R)*) but not with the feedback format weekly goal (*Aggregated*). This is what we find (Table A11).

## I NoManipulation treatment

In real life, people rarely are asked to set goals and they only get reminded of goals that they set if they write them down or take some other action to this effect. An interesting benchmark for our experiment therefore is to see how people fare in a treatment that more closely resembles the ‘natural habitat’ that subjects normally operate in. To this effect, we ran the exploratory *NoManipulation* treatment, where subjects were just informed about the task in the next week without any further manipulation of the goal bracket or any prompt to set goals. Subjects received daily emails with the link to the task.

Subjects in *NoManipulation* were not asked to set goals. Yet, the announcement of the task and time frame might have prompted subjects to privately set goals, or more broadly, to form expectations about their effort. Unless all subjects privately set daily goals in

*NoManipulation*, it follows from our theoretical results that effort should be weakly lower in *NoManipulation* than in *Daily*. To the extent that subjects in *NoManipulation* used daily goals, effort may be higher than in *Weekly*.<sup>27</sup>

Effort in *NoManipulation* did not differ significantly from the effort in *Daily* (Table A1). This suggests that when left to their own devices, subjects come close to the outcome that results when prompting them to set daily goals. The signs of the treatment dummies indicate that effort in *NoManipulation* lies between the effort in *Daily* and *Weekly*, which is consistent with a fraction of subjects *naturally* using narrowly bracketed, daily goals and the remainder either using more broadly bracketed goals or not setting any goals. These interpretations of course come with the caveat that we have no experimental control over the subjects' decisions whether and how to privately set goals. An additional confound might be that subjects adopt different goal bracketing strategies because they face different levels of uncertainty (in general or in this specific setting) and the optimal goal bracket varies with the amount of uncertainty (Koch and Nafziger, 2016). With such sorting there would be no clear cut prediction on the total effort relative to the other treatments.

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<sup>27</sup>From a theoretical point of view it is unclear what the *NoManipulation* implies for the expectations that the individual has about future effort and for comparison utility. In principle, any effort expectations (goal)  $\in [e_t^*, e_{max}]$  are self-fulfilling if the individual experiences narrowly bracketed comparison utility. If *NoManipulation* triggers no expectations or comparison utility, then we would expect that self  $t$  chooses his preferred effort  $e_t^*$ .

Table A1: Impact of eliciting goals on total effort.

Treatments	Weekly, Daily & Daily(R) vs. NoManipulation	Daily(R) vs. NoManipulation		
	(1)	(2)	(3)	(4)
Daily goals <sup>a</sup>	54.22 (105.45)	18.24 (102.58)	-27.74 (137.29)	-21.75 (130.83)
Weekly goal <sup>b</sup>	-118.54 (112.77)	-158.58 (109.13)		
Controls <sup>c</sup>	no	yes	no	yes
Margin.effect(daily goals) <sup>d</sup>	28.27	9.63	-12.76	-10.46
$\frac{\text{Margin.effect(daily goals)}}{\text{std.dev.NoManipulation(total effort)}}$	0.07	0.02	-0.03	-0.03
Margin.effect(weekly goal) <sup>d</sup>	-61.81	-83.74		
$\frac{\text{Margin.effect(weekly goal)}}{\text{std.dev.NoManipulation(total effort)}}$	-0.15	-0.21		
N	301	300	145	144

Notes. Dependent variable: Total effort. Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis.  $*p < 0.10$ ,  $**p < 0.05$ ,  $***p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Weekly*. <sup>c</sup>Controls: baseline productivity and a gender dummy. <sup>d</sup>Tobit marginal effect on the censored latent variable (at the means of control variables).

## J Gender effects

Table 3 reveals that women set significantly lower goals than men. This is consistent with the previous literature (Smithers, 2015; Dalton et al., 2015; Clark et al., 2019). Because of their lower goals, women completed fewer tables than men; but if we control for the goal level, women actually appear to have completed more tables (these results are not significant though). One reason is that women were 14 percentage points more likely to reach their goal than men according to a logit regression for the pooled sample *Daily*, *Weekly*, *Daily(R)*, and *Aggregated* ( $p = 0.016$ ,  $N = 305$ ). For some subjects we have information from a prior study (Epper et al., 2018), which contained questions on goal setting behavior and a task to measure overconfidence. Here women were less overconfident than men about their relative performance in the real effort task (permutation test,  $p < 0.001$ ,  $N = 155$ ), less likely than men to say about themselves that they set “ambitious goals” ( $p = 0.072$ ), and more often than men said that they avoided setting goals because they were afraid not to achieve them ( $p = 0.010$ ). A systematic exploration of gender differences in goal setting and achievement appears interesting for future research.

## K Supplementary tables

Table A2: Timing of treatments.

Treatment	N	Calendar week/year							
		2013				2014			
		39	40	47	48	49	50	45	48
Daily	78	37	26				15		
Weekly	77	35	25				17		
Daily(R)	75								75
Aggregated	75								75
DailyRequirement	47			26	12	9			
WeeklyRequirement	45			27	11	7			
NoManipulation	71								71

Table A3: Descriptive statistics for the additional treatments.

Treatment	N	Average total		Fraction of subjects with			Average		
		goal	effort	goal =1000	effort =1000	effort <goal	effort -goal	number of logins <sup>a</sup>	effort per login
Daily(R)	75	796 (291)	572 (427)	0.41	0.43	0.53	-224.12 (386.51)	6.32 (5.02)	99.26 (78.13)
Aggregated	75	791 (255)	649 (383)	0.48	0.43	0.39	-141.88 (407.30)	8.16 (6.11)	97.97 (71.28)
DailyRequirement	47	858 (230)	487 (447)	0.53	0.36	0.60	-370.45 (476.73)	6.32 (5.11)	83.60 (83.40)
WeeklyRequirement	45	750 (309)	558 (447)	0.53	0.42	0.40	-191.62 (483.73)	7.60 (5.82)	76.06 (38.06)

Notes. Standard deviations in parentheses. Effort: tables correctly counted.

<sup>a</sup> New login: new day or >30 min. since last entry.



Table A4: Test of balance across treatment pairs.

Treatments Variables <sup>a</sup>	Daily		Daily(R)	DailyCommitment				
	(vs. Weekly)	(vs. Aggregated)	(vs. WeeklyCommitment)	(1)	(2)	(3)	(4)	(5)
Baseline productivity	0.01	0.06	-0.05	-0.02	0.05			
Female	0.26	0.44	-0.74**	-0.10	0.92			
Self Control		-0.51			0.13			
Grit		-0.44			-0.14			
NarrowGoal		0.15			0.24			
Mentalbudget		-0.04			-0.08			
Overconfidence		0.01			0.03			
Cognitive reflection		-0.15			0.17			
High school grade		0.11			-0.67			
S-Shaped		0.71*			-0.66			
Loss aversion		0.02			0.08			
$\tau$		-1.73			-5.93			
Goal component 1		-0.08			0.15			
Goal component 2		-0.39**			0.28			
Goal component 3		0.01			0.34			
Goal component 4		0.11			-0.06			
Goal component 5		0.32**			-0.17			
Goal component 6		0.11			-0.15			
Goal component 7		-0.10			0.50*			
Constant	-0.37	1.71	1.22*	0.47	2.68			
N	155	153	150	92	78			

Notes. Dependent variable: Treatment dummy. Logit coefficient (for space reasons without standard error).

\* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup> See Appendix F for explanations of controls in (2) & (5): .

Table A5: Likelihood of achieving the total goal (logit regressions).

Treatments	Daily vs. Weekly				Daily(R) vs. Aggregated			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Daily goals <sup>a</sup>	0.02 (0.08)	0.01 (0.08)	-0.05 (0.10)	0.04 (0.08)	-0.01 (0.10)			
Daily goals feedback <sup>b</sup>						-0.15* (0.08)	-0.10 (0.09)	-0.10 (0.09)
Total goal				-.0003* (0.0001)	-.0003* (0.0002)			0.0002 (0.0002)
Baseline		0.02* (0.01)	0.05* (0.02)	0.02** (0.01)	0.05** (0.03)		0.04*** (0.01)	0.03*** (0.01)
Female		0.07 (0.08)	-0.03 (0.11)	0.01 (0.09)	-0.10 (0.11)		0.23** (0.09)	0.24*** (0.09)
Full controls <sup>c</sup>	no	no	yes	no	yes	no	no	no
N	155	155	153	155	153	150	150	150

Notes. Dependent variable: Dummy for total effort  $\geq$  total goal. Logit marginal effect with robust standard error in parenthesis. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>Full controls: see Appendix F.

Table A6: Impact of goal format on effort shortfall for non-achievers with positive effort.

Treatments	Daily vs. Weekly			Daily(R) vs. Aggregated	
	(1)	(2)	(3)	(4)	(5)
Daily goals <sup>a</sup>	112.98 (72.29)	126.60* (65.86)	134.77** (63.56)		
Daily goals feedback <sup>b</sup>				56.71 (80.79)	3.98 (75.08)
Baseline productivity		-9.09 (9.13)	-18.24 (16.55)		-26.44*** (9.73)
Female		272.31*** (66.98)	309.92*** (85.70)		-12.34 (82.34)
Constant	-461.58*** (49.83)	-452.80*** (134.51)	155.85 (373.16)	-532.58*** (53.44)	-91.19 (170.82)
Full controls <sup>b</sup>	no	no	yes	no	no
Margin.effect(daily goals) <sup>c</sup>	101.43	117.72*	128.71**	50.76	3.62
N	65	65	65	56	56

Notes. Dependent variable: Total effort-total goal. Sample: non-achievers with positive effort ( $0 < total\ effort < total\ goal$ ). Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>See Appendix F. <sup>d</sup>Tobit marginal effect on the censored latent variable (at the means of control variables).

Table A7: Profiles of goals (*Daily*).

Profile	Goal achiever		Non-achiever		All		Achievement rate
	N	Percent	N	Percent	N	Percent	
<b>‘Flat’</b> ( $g_{mon} = g_{tue} = \dots = g_{fri}$ )	<b>9</b>	<b>20.93</b>	<b>8</b>	<b>22.86</b>	<b>17</b>	<b>21.79</b>	<b>52.94</b>
Daily goal = 200	6	13.95	5	14.29	11	14.10	54.55
Daily goal = 100	1	2.33	0	0.00	1	1.28	100
Other	2	4.65	3	8.57	5	6.41	40.00
<b>‘High-low’</b> ( $(g_{mon} + g_{tue})/2 > (g_{thu} + g_{fri})/2$ )	<b>14</b>	<b>32.56</b>	<b>17</b>	<b>48.57</b>	<b>31</b>	<b>39.74</b>	<b>45.16</b>
$(g_{mon} + g_{tue})/2 \geq g_{wed} \geq (g_{thu} + g_{fri})/2$ <sup>a</sup>	6	13.95	7	20.00	13	16.67	46.15
Decreasing <sup>b</sup>	4	9.30	4	11.43	8	10.26	50.00
One daily goal = 1000 <sup>c</sup>	1	2.33	0	0.00	1	1.28	100
<b>‘Low-high’</b> ( $(g_{mon} + g_{tue})/2 < (g_{thu} + g_{fri})/2$ )	<b>18</b>	<b>41.86</b>	<b>9</b>	<b>25.71</b>	<b>27</b>	<b>34.62</b>	<b>66.67</b>
$(g_{mon} + g_{tue})/2 \leq g_{wed} \leq (g_{thu} + g_{fri})/2$ <sup>a</sup>	8	18.60	7	20.00	15	19.23	53.33
Increasing <sup>d</sup>	3	6.98	1	2.86	4	5.13	75.00
<b>Other type of profile</b>	<b>2</b>	<b>4.65</b>	<b>1</b>	<b>2.86</b>	<b>3</b>	<b>3.85</b>	<b>66.67</b>
<b>All</b>	<b>43</b>	<b>100</b>	<b>35</b>	<b>100</b>	<b>78</b>	<b>100</b>	<b>55.13</b>
All daily goals > 0	39	90.70	28	80.00	67	85.90	58.21
At least one daily goal = 0 <sup>e</sup>	3	6.98	7	20.00	10	12.82	30.00
$g_{mon} = 0$	0	0.00	1	2.86	1	1.28	0
$g_{tue} = 0$	0	0.00	4	11.43	4	5.13	0
$g_{wed} = 0$	1	2.33	1	2.86	2	2.56	50.00
$g_{thu} = 0$	1	2.33	0	0.00	1	1.28	100
$g_{fri} = 0$	1	2.33	5	14.29	6	7.69	16.67

Notes. <sup>a</sup>At least one inequality strict. <sup>b</sup> $g_{mon} \geq g_{tue} \geq g_{wed} \geq g_{thu} \geq g_{fri}$ , at least one inequality strict. <sup>c</sup>Here,  $g_{mon} = 1000$ . <sup>d</sup> $g_{mon} \leq g_{tue} \leq g_{wed} \leq g_{thu} \leq g_{fri}$ , at least one inequality strict. <sup>e</sup>Excludes the case where because  $g_{mon} = 1000$ , mechanically, the other daily goals are zero. Some subjects had more than one day with a zero goal.

Table A8: Profiles of goals (all treatments with daily goals).

Profile	N	Percent
<b>‘Flat’</b> ( $g_{mon} = g_{tue} = \dots = g_{fri}$ )	<b>47</b>	<b>20.61</b>
Daily goal = 200	26	11.40
Daily goal = 100	9	3.95
Other	12	5.26
<b>‘High-low’</b> ( $(g_{mon} + g_{tue})/2 > (g_{thu} + g_{fri})/2$ )	<b>100</b>	<b>43.86</b>
$(g_{mon} + g_{tue})/2 \geq g_{wed} \geq (g_{thu} + g_{fri})/2$ <sup>a</sup>	45	19.74
Decreasing <sup>b</sup>	19	8.33
One daily goal = 1000 <sup>c</sup>	2	0.88
<b>‘Low-high’</b> ( $(g_{mon} + g_{tue})/2 < (g_{thu} + g_{fri})/2$ )	<b>71</b>	<b>31.14</b>
$(g_{mon} + g_{tue})/2 \leq g_{wed} \leq (g_{thu} + g_{fri})/2$ <sup>a</sup>	30	13.16
Increasing <sup>d</sup>	10	4.39
<b>Other type of profile</b>	<b>10</b>	<b>4.39</b>
<b>All</b>	<b>228</b>	<b>100</b>
All daily goals > 0	195	85.53
At least one daily goal = 0 <sup>e</sup>	31	13.60
$g_{mon} = 0$	7	3.07
$g_{tue} = 0$	7	3.07
$g_{wed} = 0$	6	2.63
$g_{thu} = 0$	10	4.39
$g_{fri} = 0$	17	7.46

Notes. Includes all treatments where subjects set daily goals with the same instructions: *Daily*, *Daily(R)*, and *Aggregated*. <sup>a</sup>At least one inequality strict. <sup>b</sup> $g_{mon} \geq g_{tue} \geq g_{wed} \geq g_{thu} \geq g_{fri}$ , at least one inequality strict. <sup>c</sup>In both cases,  $g_{mon} = 1000$ . <sup>d</sup> $g_{mon} \leq g_{tue} \leq g_{wed} \leq g_{thu} \leq g_{fri}$ , at least one inequality strict. <sup>e</sup>Excludes the case where because  $g_{mon} = 1000$ , mechanically, the other daily goals are zero. Some subjects had more than one day with a zero goal.

Table A9: Transition from goal profiles to effort profiles in *Daily*.

Goal profiles		Effort profiles							
		Flat <sup>d</sup>		High-low <sup>e</sup>		Low-high <sup>f</sup>		Other	
		N	Percent	N	Percent	N	Percent	N	Percent
Goal achievers <sup>g</sup>	Flat <sup>a</sup>	1	11.11	5	55.56	3	33.33	0	0.00
	High-low <sup>b</sup>	0	0.00	12	85.71	2	14.29	0	0.00
	Low-high <sup>c</sup>	0	0.00	9	50.00	9	50.00	0	0.00
	Other	0	0.00	1	50.00	1	50.00	0	0.00
Goal non-achievers <sup>g</sup>	Flat <sup>a</sup>	0	0.00	7	87.50	0	0.00	1	12.50
	High-low <sup>b</sup>	0	0.00	12	70.59	3	17.65	2	11.76
	Low-high <sup>c</sup>	0	0.00	4	44.44	4	44.44	1	11.11
	Other	0	0.00	1	100.00	0	0.00	0	0.00

Notes. <sup>a</sup>  $g_{mon} = g_{tue} = \dots = g_{fri}$ . <sup>b</sup>  $g_{mon} + g_{tue} > g_{thu} + g_{fri}$ . <sup>c</sup>  $g_{mon} + g_{tue} < g_{thu} + g_{fri}$ .

<sup>d,e,f</sup> Analogous to goal profiles. <sup>g(h)</sup> Total effort > (<=) total goal.

Table A10: Profiles of effort (*Daily* and *Weekly*).

Profile	Daily				Weekly			
	Goal achiever		Non-achiever		Goal achiever		Non-achiever	
	N	Percent	N	Percent	N	Percent	N	Percent
<b>‘Flat’</b> ( $e_{mon} = e_{tue} = \dots = e_{fri}$ )	<b>1</b>	<b>2.33</b>	<b>0</b>	<b>0.00</b>	<b>0</b>	<b>0.00</b>	<b>0</b>	<b>0.00</b>
Daily effort = 200	1	2.33	0	0.00	0	0.00	0	0.00
<b>‘High-low’</b> ( $(e_{mon} + e_{tue})/2 > (e_{thu} + e_{fri})/2$ )	<b>27</b>	<b>62.79</b>	<b>24</b>	<b>68.57</b>	<b>25</b>	<b>60.98</b>	<b>21</b>	<b>58.33</b>
$(e_{mon} + e_{tue})/2 \geq e_{wed} \geq (e_{thu} + e_{fri})/2$ <sup>a</sup>	12	27.91	14	40.00	13	31.71	12	33.33
Decreasing <sup>b</sup>	7	16.28	11	31.43	5	12.20	7	19.44
One daily effort = 1000 <sup>c</sup>	4	9.30	0	0.00	3	7.32	0	0.00
<b>‘Low-high’</b> ( $(e_{mon} + e_{tue})/2 < (e_{thu} + e_{fri})/2$ )	<b>15</b>	<b>34.88</b>	<b>7</b>	<b>20.00</b>	<b>14</b>	<b>34.15</b>	<b>11</b>	<b>30.56</b>
$(e_{mon} + e_{tue})/2 \leq e_{wed} \leq (e_{thu} + e_{fri})/2$ <sup>a</sup>	4	9.30	2	5.71	5	12.20	5	13.89
Increasing <sup>d</sup>	1	2.33	0	0.00	0	0.00	4	11.11
One daily effort = 1000	0	0.00	0	0.00	0	0.00	0	0.00
<b>Total effort = 0</b>	<b>0</b>	<b>0.00</b>	<b>3</b>	<b>8.57</b>	<b>0</b>	<b>0.00</b>	<b>3</b>	<b>8.33</b>
<b>Other type of profile</b>	<b>0</b>	<b>0.00</b>	<b>1</b>	<b>2.86</b>	<b>2</b>	<b>4.88</b>	<b>1</b>	<b>2.78</b>
<b>All</b>	<b>43</b>	<b>100</b>	<b>35</b>	<b>100</b>	<b>41</b>	<b>100</b>	<b>36</b>	<b>100</b>
All daily effort levels > 0	15	34.88	9	25.71	4	9.76	0	0.00
At least one daily effort = 0 <sup>e</sup>	24	55.81	26	74.29	33	80.49	36	100.00
$e_{mon} = 0$	6	13.95	10	28.57	7	17.07	18	50.00
$e_{tue} = 0$	3	6.98	15	42.86	6	14.63	20	55.56
$e_{wed} = 0$	11	25.58	13	37.14	13	31.71	22	61.11
$e_{thu} = 0$	15	34.88	18	51.43	16	39.02	21	58.33
$e_{fri} = 0$	14	32.56	21	60.00	21	51.22	27	75.00
Average total goal	736.53		852.37		651.37		716.56	
Average total effort	886.88		448.06		774.17		232.31	
Baseline productivity	17.09		16.37		17.32		15.42	

Notes. <sup>a</sup>At least one inequality strict. <sup>b</sup> $e_{mon} \geq e_{tue} \geq e_{wed} \geq e_{thu} \geq e_{fri}$ , at least one inequality strict. <sup>c</sup>In all cases,  $e_{mon} = 1000$ . <sup>d</sup> $e_{mon} \leq e_{tue} \leq e_{wed} \leq e_{thu} \leq e_{fri}$ , at least one inequality strict. <sup>e</sup>Excludes the cases where because  $e_{mon} = 1000$  or  $e_{wed} = 1000$ , mechanically, the other daily goals are zero. Some subjects had more than one day with a zero effort.

Table A11: Check of successful framing (*Daily(R)* vs. *Aggregated*).

Dependent variable	Daily effort (tables counted)				
	Mon	Tue	Wed	Thu	Fri
(Daily goal level)	0.74***	0.98***	0.87**	0.63**	0.72
x Daily <sup>a</sup>	(0.26)	(0.23)	(0.34)	(0.28)	(0.45)
(Daily goal level)	0.11	0.21	0.08	-0.30	0.61*
x Aggregated <sup>a</sup>	(0.16)	(0.29)	(0.28)	(0.36)	(0.34)
N	150	149 <sup>b</sup>	139 <sup>b</sup>	133 <sup>b</sup>	120 <sup>b</sup>

Notes. Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis. All models include a dummy for *Daily(R)*, baseline productivity, a gender dummy, and a constant. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Interaction between treatment dummy and the daily goal level. <sup>b</sup>Excludes subjects whose cumulative completed tables reached 1000 on the previous day.



Table A12: Impact of goal feedback format on total effort (*Daily(R)* vs. *Aggregated*).

	(1)	(2)	(3)
Daily goals feedback <sup>a</sup>	-108.26 (132.05)	-17.45 (130.91)	-37.37 (122.85)
Total goal			1.07*** (0.19)
Baseline productivity		51.72*** (16.73)	38.69** (16.03)
Female		247.14* (132.17)	299.74** (124.93)
Constant	856.74*** (96.52)	-166.93 (311.69)	-823.58*** (309.98)
Margin.effect(daily goals feedback) <sup>b</sup>	-50.22	-8.40	-19.09
Effect size <sup>c</sup>	-0.13	-0.02	-0.05
N	150	150	150

Notes. Dependent variable: Total effort. Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily(R)*. <sup>b</sup>Tobit marginal effect on the censored latent variable (at the means of control variables).

<sup>c</sup> $\frac{\text{Margin.effect(daily goals feedback)}}{\text{Standard deviation of total goal in } \textit{Aggregated}}$ .

Table A13: Impact of goal and goal setting or feedback format on effort on Monday.

Treatments	Daily vs. Weekly				Daily(R) vs. Aggregated				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8) <sup>f</sup>	(9)
Daily goals <sup>a</sup>	85.48 (56.14)	93.47* (56.10)	102.05* (57.75)	65.99 (56.10)	73.29 (56.54)				
Daily goals feedback <sup>b</sup>						63.75 (42.75)	90.14** (41.09)	87.49** (39.09)	88.35** (39.22)
Total goal				0.24*** (0.08)	0.23*** (0.08)			0.30*** (0.07)	0.28*** (0.08)
Monday goal									0.05 (0.17)
Constant	93.74** (38.75)	13.67 (140.91)	-71.67 (373.15)	-122.34 (145.04)	-186.65 (371.42)	105.91*** (28.30)	-250.45*** (94.33)	-435.18*** (100.28)	-434.70*** (100.87)
Margin.effect(daily goals) <sup>d</sup>	56.01	61.62*	68.63*	43.90	49.81	45.20	64.88**	63.75**	64.40**
Effect size <sup>e</sup>	0.24	0.24	0.29	0.19	0.21	0.27	0.38	0.38	0.38
Controls <sup>c</sup>	no	yes	full	yes	full	no	yes	yes	yes
N	155	155	153	155	153	150	150	150	150

Notes. Dependent variable: Tables counted on Monday. Tobit coefficient (marginal effect on the latent dependent variable). \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>baseline productivity and a gender dummy; full controls: see Appendix F. <sup>d</sup>Tobit marginal effect on the censored latent variable (at the means of control variables). <sup>e</sup>Standard deviation of effort on Monday in *Weekly* or *Aggregated*, respectively. <sup>f</sup>Replacing *total goal* with the weekdays  $g_{mon}, \dots, g_{fri}$  yields treatment effect 98.93 (se 42.56,  $p = 0.022$ ) and marginal effect 72.2.

Table A14: Impact of goal and goal setting or feedback format on effort on Friday.

Treatments	Daily vs. Weekly				Daily(R) vs. Aggregated				
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8) <sup>f</sup>	(9)
Daily goals <sup>a</sup>	110.48*	102.60*	127.03**	96.51*	122.57**				
	(56.09)	(55.52)	(56.68)	(56.84)	(57.21)				
Daily goals feedback <sup>b</sup>						-120.16***	-98.36**	-103.96**	-111.01***
						(42.93)	(42.05)	(40.60)	(42.15)
Total goal				0.05	0.03			0.22***	0.17*
				(0.09)	(0.09)			(0.08)	(0.10)
Friday goal									0.34
									(0.35)
Constant	-128.55***	-189.19*	-268.07	-217.46*	-283.49	53.37*	-44.66	-191.54*	-184.54*
	(47.07)	(107.35)	(332.31)	(121.26)	(332.18)	(29.47)	(96.55)	(105.90)	(105.51)
Margin.effect(daily goals) <sup>d</sup>	44.65	41.23*	48.95*	38.81	47.22	-58.71	-48.12**	-50.50**	-53.93**
Effect size <sup>e</sup>	0.30	0.30	0.32	0.26	0.31	-0.41	-0.33	-0.35	-0.37
Controls <sup>c</sup>	no	yes	full	yes	full	no	yes	yes	yes
N	155	155	153	155	153	150	150	150	150

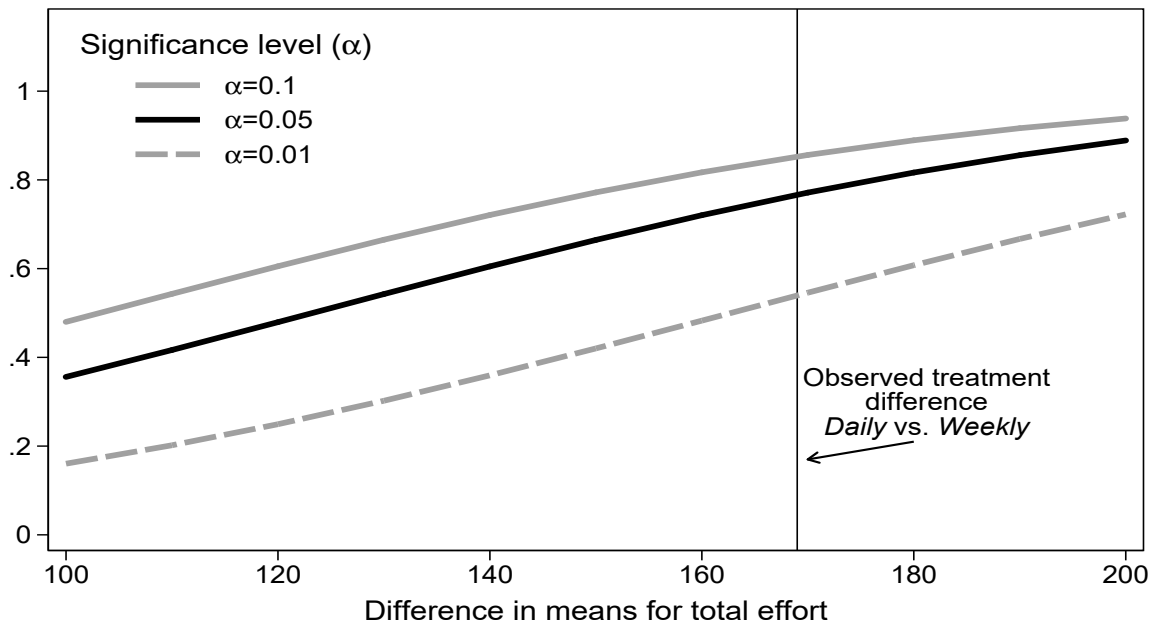
Notes. Dependent variable: Tables counted on Friday. Tobit coefficient (marginal effect on the latent dependent variable). \*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ . <sup>a</sup>Dummy for *Daily(R)*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>baseline productivity and a gender dummy; full controls: see Appendix F. <sup>d</sup>Tobit marginal effect on the censored latent variable (at the means of control variables). <sup>e</sup>Standard deviation of effort on Friday in *Weekly* or *Aggregated*, respectively. <sup>f</sup>Replacing *total goal* with the weekdays  $g_{mon}, \dots, g_{fri}$  yields treatment effect -110.21 (se 42.56,  $p = 0.011$ ) and marginal effect -53.62.

Table A15: Impact of goal setting format on total goal, total effort, and effort on Monday (*DailyRequirement* vs. *WeeklyRequirement*).

Dependent variable	Total goal	Total effort		Effort Monday	
	(1)	(2)	(3)	(4)	(5)
Daily goals <sup>a</sup>	135.31 (114.22)	-106.68 (269.39)	-186.19 (272.26)	43.76 (55.95)	30.84 (54.47)
Total goal			0.70 (0.44)		0.12 (0.09)
Baseline productivity	5.65 (14.69)	25.61 (32.44)	23.48 (31.93)	3.66 (6.21)	3.42 (6.27)
Female	-108.91 (121.35)	-45.51 (291.04)	-23.74 (286.54)	0.23 (61.40)	3.29 (60.17)
Constant	926.63*** (264.44)	315.76 (599.69)	-179.66 (610.58)	53.34 (110.01)	-37.31 (112.63)
Margin.effect(daily goals) <sup>b</sup>	69.59	-35.54	-62.62	29.17	20.94
Effect size <sup>c</sup>	0.25	-0.08	-0.14	0.07	0.05
N	92	92	92	92	92

Notes. Dependent variable: Total goal in (1)&(2), total effort (tables counted over all five days) in (3)&(4), and tables counted on Monday in (5)&(6). Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis. Results with the full set of controls are similar. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *DailyRequirement*. <sup>b</sup>Tobit marginal effect on the censored latent variable (at the means of control variables). <sup>c</sup> $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of outcome in } \textit{WeeklyRequirement}}$

Figure A4: Power analysis (two-sided test of  $H_0$  no mean difference).



Parameters:  $N = 155$ ,  $N_1 = 78$ ,  $N_2 = 77$ ,  $\mu_1 = 521$ ,  $\sigma = 389$

## L Robustness

The tobit model yields consistent and efficient estimates if errors are normally distributed and homoscedastic. To assess the robustness of our findings, we report in tables A16 -A19 results from ordinary least squares (OLS) estimation and two semiparametric estimators that are robust to specification problems of the tobit model:

- (i) The OLS coefficients have the direct interpretation as marginal effect on the expected value of the dependent variable.
- (ii) The censored least absolute deviations (CLAD) estimator (Powell, 1984) permits non-normal, heteroskedastic, and asymmetric errors because it is consistent as long as errors have a median of zero.
- (iii) The symmetrically censored least squares (SCLS) estimator (Powell, 1986) requires that the errors are symmetrically distributed around zero (a stronger condition than the zero median restriction of the CLAD estimator).

Computation of the CLAD and SCLS estimators involves iterative procedures that delete some observations, such that the number of effectively used observations is smaller than the initial sample (cf. Chay and Powell, 2001). In the case of CLAD, the procedure alternates between deleting observations for which the current regression function yields estimates that fall outside of the uncensored region and applying least absolute deviations estimation to the remaining observations. In the case of SCLS, the procedure corrects for the censoring at the top by ‘symmetrically censoring’ from below so that the ‘re-censored’ dependent variable is symmetrically distributed around the regression function, and it drops observations for which the current estimates are outside of the uncensored region.

The CLAD and SCLS estimates can be compared with the tobit coefficients, whereas the OLS estimates should be compared with the tobit marginal effect on the expected value of the censored dependent variable. Overall, we see that the tobit estimator, if at all, understates treatment effects.

Table A16: Impact of goal setting format or goal feedback format on total goal.

Treatments	Daily vs. Weekly			Daily(R) vs. Aggregated				
	OLS (1)	Tobit (2)	CLAD (3)	SCLS (4)	OLS (5)	Tobit (6)	CLAD (7)	SCLS (8)
Daily goals <sup>a</sup>	118.19** (48.45)	173.14* (89.37)	193.40 (146.92)	399.84** (189.39)				
Daily goals feedback <sup>b</sup>					7.90 (43.94)	-15.52 (75.77)	46.80 (109.56)	160.56 (182.82)
Productivity	12.92** (5.90)	24.17** (11.08)	34.07* (21.05)	55.14 (51.83)	11.59** (4.61)	25.79*** (9.01)	7.80* (9.35)	14.34 (30.57)
Female	-226.08*** (47.14)	-417.57*** (90.65)	-465.93*** (146.18)	-429.58 (337.20)	-34.67 (44.96)	-87.34 (78.00)	-39.00 (106.46)	-162.00 (424.92)
Constant	584.04*** (101.80)	677.16*** (184.81)	489.00 (323.91)	-132.26 (88.12)	624.15*** (90.37)	587.68*** (160.16)	844.00*** (183.97)	-220.69 (178.41)
Margin.effect(daily goals) <sup>c</sup>	118.19*	89.44**			7.90	-8.41		
Effect size <sup>d</sup>	0.35	0.27			0.03	-0.03		
N	155	155	155(105) <sup>e</sup>	155(88) <sup>e</sup>	150	150	150(117) <sup>e</sup>	150(85) <sup>e</sup>

Notes. Dependent variable: Total goal. Coefficient for ordinary least squares (OLS); tobit coefficient (marginal effect on the latent dependent variable), censored least absolute deviations estimator (CLAD), and symmetrically censored least squares (SCLS). Robust standard error in parenthesis for OLS, tobit, and SCLS. For CLAD, bootstrap standard error in parenthesis and the significance level is based on bias-corrected bootstrap confidence intervals. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>OLS coefficient/tobit marginal effect on the censored latent variable (at the means of control variables). <sup>d</sup>  $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly or Aggregated}}$ , respectively. <sup>e</sup>Number of observations effectively used in the CLAD/SCLS estimation in parenthesis.

Table A17: Impact of goal and goal setting or feedback format on total effort (without control for total goal).

Treatments	Daily vs. Weekly			Daily(R) vs. Aggregated				
	OLS	Tobit	CLAD	SCLS	OLS	Tobit	CLAD	SCLS
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Daily goals <sup>a</sup>	164.03*** (60.35)	228.39** (94.86)	239.08 (131.49)	638.10*** (126.45)				
Daily goals feedback <sup>b</sup>					-37.92 (65.82)	-17.45 (130.91)	-84.00 (184.71)	-134.00 (195.55)
Productivity	20.80*** (6.27)	33.52*** (10.67)	32.62*** (14.94)	46.78** (20.22)	22.90*** (7.25)	51.72*** (16.73)	42.00** (18.05)	45.05 (31.98)
Female	-27.95 (62.21)	-93.66 (99.86)	-75.92 (122.62)	-142.22 (179.17)	126.73* (68.50)	247.14* (132.17)	294.00 (180.96)	145.07 (194.30)
Constant	193.35* (110.24)	109.27 (176.97)	43.38 (264.03)	-131.38*** (27.85)	186.90 (140.28)	-166.93 (311.69)	-176.00 (325.98)	-3.85** (1.90)
Margin.effect(daily goals) <sup>c</sup>	164.03***	134.38**			-37.92	-8.40		
Effect size <sup>d</sup>	0.42	0.34			-0.10	-0.02		
N	155	155	155(150) <sup>e</sup>	155(88) <sup>e</sup>	150	150	150(144) <sup>e</sup>	150(134) <sup>e</sup>

Notes. Dependent variable: Total effort. Coefficient for ordinary least squares (OLS); tobit coefficient (marginal effect on the latent dependent variable), censored least absolute deviations estimator (CLAD), and symmetrically censored least squares (SCLS). Robust standard error in parenthesis for OLS, tobit, and SCLS. For CLAD, bootstrap standard error in parenthesis and the significance level is based on bias-corrected bootstrap confidence intervals. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>OLS coefficient/tobit marginal effect on the censored latent variable (at the means of control variables). <sup>d</sup>  $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly or Aggregated}}$ , respectively. <sup>e</sup>Number of observations effectively used in the CLAD/SCLS estimation in parenthesis.



Table A18: Impact of goal and goal setting or feedback format on total effort (controlling for total goal).

Treatments	Daily vs. Weekly			Daily(R) vs. Aggregated				
	OLS (1)	Tobit (2)	CLAD (3)	SCLS (4)	OLS (5)	Tobit (6)	CLAD (7)	SCLS (8)
Daily goals <sup>a</sup>	111.14* (58.18)	151.44* (91.31)	187.08** (99.73)	201.96 (124.89)				
Daily goals feedback <sup>b</sup>					-42.01 (62.18)	-37.37 (122.85)	-0.00 (127.21)	-66.60 (132.80)
Total goal	0.45*** (0.09)	0.60*** (0.14)	0.82*** (0.16)	0.68*** (0.17)	0.52*** (0.10)	1.07*** (0.19)	0.89*** (0.13)	0.84*** (0.21)
Productivity	15.01** (6.17)	24.49** (10.38)	18.60* (12.09)	18.92 (11.99)	16.89** (7.29)	38.69** (16.03)	0.00 (19.31)	29.15** (14.59)
Female	73.22 (61.24)	53.56 (99.11)	40.46 (108.76)	48.54 (121.68)	144.71** (65.13)	299.74** (124.93)	249.00 (163.20)	147.26 (136.01)
Constant	-68.01 (109.22)	-230.05 (168.23)	-319.90 (189.61)	-208.67*** (30.84)	-136.65 (140.22)	-823.58*** (309.98)	-141.77 (321.17)	-426.54*** (90.94)
Margin.effect(daily goals) <sup>c</sup>	111.14* 0.28	93.93* 0.24			-42.01 -0.11	-19.09 -0.05		
N	155	155	155(131) <sup>e</sup>	155(129) <sup>e</sup>	150	150	150(150) <sup>e</sup>	150(124) <sup>e</sup>

Notes. Dependent variable: Total effort. Coefficient for ordinary least squares (OLS); tobit coefficient (marginal effect on the latent dependent variable), censored least absolute deviations estimator (CLAD), and symmetrically censored least squares (SCLS). Robust standard error in parenthesis for OLS, tobit, and SCLS. For CLAD, bootstrap standard error in parenthesis and the significance level is based on bias-corrected bootstrap confidence intervals. \* $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>OLS coefficient/tobit marginal effect on the censored latent variable (at the means of control variables). <sup>d</sup> $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly or Aggregated}}$ . <sup>e</sup>Number of observations effectively used in the CLAD/SCLS estimation in parenthesis.

Table A19: Getting started: Effort on Monday (OLS regressions).

Treatments	Daily vs.		DailyRequirement				Daily(R) vs.			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
	Weekly		WeeklyRequirement				Aggregated			
Daily goals <sup>a</sup>	51.71 (39.53)	54.78 (42.80)	30.96 (40.18)	32.15 (42.46)	33.93 (41.97)	64.21 (50.29)	20.98 (40.97)	42.44 (49.25)		
Daily goals feedback <sup>b</sup>									64.39** (30.68)	62.81** (29.51)
Total goal			0.18*** (0.05)	0.18*** (0.06)			0.12* (0.07)	0.18* (0.09)		0.20*** (0.04)
Baseline	6.77 (5.95)	21.12* (12.66)	4.50 (5.69)	18.78 (12.27)	2.86 (4.70)	3.74 (11.48)	2.65 (4.83)	7.25 (12.14)	14.38*** (3.86)	12.05*** (3.77)
Female	-83.48** (39.44)	-72.74 (46.85)	-43.78 (38.48)	-31.59 (46.20)	-15.59 (47.30)	-60.69 (63.56)	-11.62 (46.40)	-49.93 (63.65)	43.31 (31.31)	50.27* (30.16)
Constant	99.40 (100.51)	78.03 (274.46)	-3.15 (101.14)	-20.58 (273.94)	125.74 (83.40)	-21.46 (385.84)	36.28 (81.74)	-175.85 (397.96)	-111.18* (66.05)	-236.36*** (64.78)
Full controls <sup>c</sup>	no	yes	no	yes	no	yes	no	yes	no	no
Effect size <sup>d</sup>	0.22	0.23	0.23	0.14	0.17	0.36	0.09	0.24	0.38	0.37
N	155	153	155	153	92	78	92	78	150	150

Notes. Dependent variable: Tables counted on Monday. OLS coefficient with robust standard error in parenthesis. (We do not report CLAD/SCLS estimates because only 8 out of the 397 subjects in the treatments with goal setting counted 1000 tables on Monday).  $p < 0.10$ , \*\* $p < 0.05$ , \*\*\* $p < 0.01$ . <sup>a</sup>Dummy for *Daily* or *DailyRequirement*. <sup>b</sup>Dummy for *Daily(R)*. <sup>c</sup>Full controls: see Appendix F. <sup>d</sup>Standard deviation of effort on Monday in Treatment *Weekly*, *WeeklyRequirement*, *Aggregated* or *Weekly*, respectively.

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# M Instructions<sup>1</sup>

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[Participants can choose on each screen between English and Danish – below are the English instructions]

## Week 1 (Goal setting)

### Screen 1

Welcome to the third part of the scientific study on Aarhus University students' traits, behaviors and study outcomes.

**By participating you can earn up to 500 kr.**

**Your tasks:** Next week from Monday, [date] - 0:00h until Friday, [date] - 23:59h you have the opportunity to count in total up to 1000 tables – just like the tables you counted in the previous parts of this study. **You earn 50 øre for each table where you count the number of zeros correctly.** If you miscount a table, you will be asked to count it again.

**Show an example table (click here).**

Tables look like follows and once you have counted the number of zeros in a table, you should enter the number of zeros in that table into a field below the table.

1	0	0	1	1
0	0	1	0	1
0	0	0	0	1
1	1	0	1	1
0	0	1	0	1
0	0	0	0	1

How many zeros are in the table?

(17 is the correct answer for this table)

### Close window

Each day at 0:00h you will receive an email with a personal link that allows you to log in and count tables. **You can count as many of the 1000 tables as you like.** Your answers will be automatically saved when you move

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<sup>1</sup> Instructions shown are the ones administered through a larger online study, for which participants were recruited through an email call to all first-year students at the School of Business and Social Sciences. Instructions were subjects were recruited over the Cognition and Behavior Lab are analogous.

to a new screen, and you can use your personalized link from the email to return as often as you like from Monday, 30.09. - 0:00h until Friday, 04.10. - 23:59h.

**To participate, you now have to complete the next two screens by setting goals for how much you want to work next week.**

**Payments:** Like for the first parts of the study, Aarhus University will automatically transfer the amount you earn into your [NemKonto](#). Alexander Koch and his team will start registering the payments with the administration of Aarhus University in week X ([date]-[date]). Then the administrative process might take between 2-6 weeks. You can contact Alexander Koch by email ([akoch@econ.au.dk](mailto:akoch@econ.au.dk)) if you want information on the payment process.

**Taxes:** According to Danish law, Aarhus University reports payments to the tax authorities. Please note that taxes might be deducted from the amount of money you earn. That is, the amount you will receive might be lower than the one stated.

- Yes, I want to participate.
- No, thanks.

### Screen 1 (Treatments DailyStart and WeeklyStart)

Same as above, just extra text on screen 1:

The only condition to be eligible to receive any payment (up to 500 kr.) is to log on at least once a day (Monday to Friday) and to correctly count at least one table every day. You get no payment for the first table in a day that you correctly count.

### Screen 2

Next week you have the opportunity to count in total up to 1000 tables. **You earn 50 øre for each table where you count the number of zeros correctly. So all in all, you can earn up to 500 kr.**

**On the next page you will set yourself a goal for how much you want to work next week.**

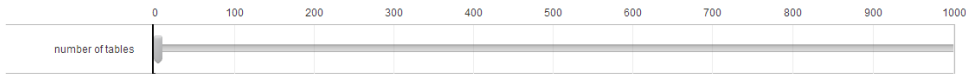
Before doing so, please take a moment to think about the following questions:

**How much time do you think you have next week to work on this task?**



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**How many tables do you think you can realistically manage to solve within that time?**



**How much money would you like to earn?**



**What you would like to do with the money that you earn over the next week?**

**Please write a short description here:**

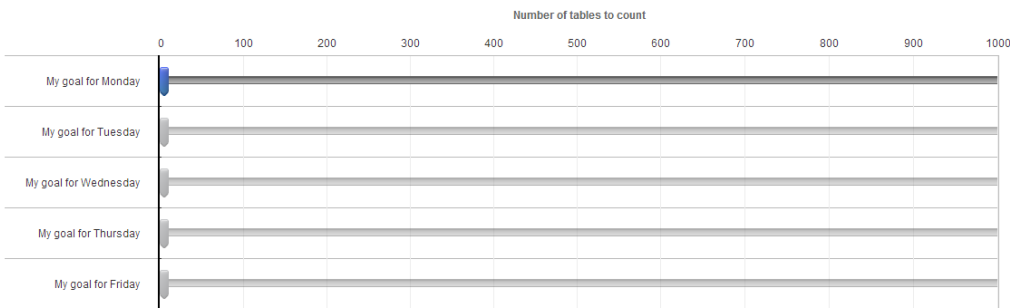
### Screen 3 (Treatment Daily, Daily14, DailyAggregated)

#### Set goals!

**Now set yourself a goal for how many tables to count on each weekday.** Next week we will then remind you of your goals. But, of course, you are free to work as much as you want.

Remember:

- You can log in as often as you like with the personal link that you will receive in an email and count tables anytime from Monday, 30.09. - 0:00h until Friday, 04.10. - 23:59h.
- You can count up to 1000 tables in total over the five days.
- You earn 50 øre for each table where you count the number of zeros correctly.

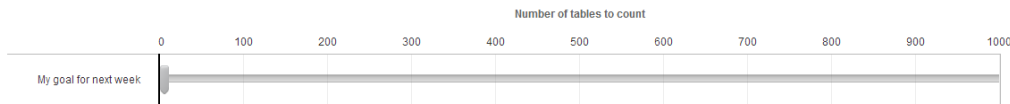


### Screen 3 (Treatment Weekly)

#### Set a goal!

**Now set yourself a goal for how many tables to count next week.** Next week we will then remind you of your goal. But, of course, you are free to work as much as you want.

Remember: [as above]



### Screen 3 (Treatments DailyStart and WeeklyStart)

Analogue to screens 3 above, but add last bullet to “Remember”:

- The only condition to be eligible to receive any payment (up to 500 kr.) is to log on at least once a day (Monday to Friday) and to correctly count at least one table every day.

### Screen 2 and 3 (Treatments NoGoal)

[These screens do not appear in this treatment.]

[As this treatment was not part of the larger online survey, written informed consent is given at this stage]

### Screen 4

Thanks!

Your answers have been registered. Monday, 30.09. 0:00h you will receive an email with a personal link that allows you to log in and start counting tables.

Finish and Save

## Week 2 (Work task)

### Screen (Treatment Daily, Daily14)

**Your goal for today: count [goal] tables.**

**So far you counted [counted tables] tables today.**

You can count as many of the remaining [remaining tables] tables as you like. **You earn 50 øre for each table where you count the number of zeros correctly.**

**Please count the number of zeros in the following table.** Once you counted the table, please click “>>” to save your response. If you miscount the table, you will be asked to count it again. If you want to stop counting simply close the browser. You can continue counting until Friday, [date] - 23:59h by logging in with the personal link from the email you received.

[Table]

How many zeros are in the table?

### Screen (Treatment Weekly)

**Your goal for this week: count [goal] tables until Friday 23:59h.**

**So far you counted [counted tables] tables this week.**

[Rest as above]

**Screen (Treatment DailyAggregated)**

[As for treatment Weekly]

**Screen (Treatment NoGoal)**

[As above. The line "Your goal for this week (...)" is deleted]

**Screen (Treatment DailyStart, WeeklyStart)**

Same as above, just extra text:

Remember that the condition to be eligible to receive any payment (up to 500 kr.) is to log on at least once a day (Monday to Friday) and to correctly count at least one table every day.



# Email texts<sup>2</sup>

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All emails had the following structures (below we only state the english versions of the main body):

**Subject:** [First Name]: Deltagelse i 3. del af den videnskabelige undersøgelse på Aarhus Universitet /  
Participation in 3rd part of the scientific study at Aarhus University

*For an English version please see below*

[Danish version]

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[English version]

## **Invitation (Wednesday 00:00, week 1)**

Dear [First Name Last Name],

Last week, you agreed to participate in the third part of the scientific study on students' traits, behaviors and study outcomes. **By participating you can earn up to 500 kr.**

To get started please click on the following link (or copy it into your internet browser) before 23:59h on Friday, [date]:

[Link]

Use a desktop computer, notebook or an iPad to participate in this study. Unfortunately, it is not possible to use a smartphone (such as an iPhone or BlackBerry).

Many thanks for participating in this study,  
Alexander Koch (Institut for Økonomi, Aarhus Universitet)

## **Reminder (if incomplete, Friday 9:00, week 1)**

Dear [First Name Last Name],

Last week, you agreed to participate in the third part of the scientific study on students' traits, behaviors and study outcomes. **By participating you can earn up to 500 kr.**

To get started please click on the following link (or copy it into your internet browser) before 23:59h tonight (Friday, [date]):

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<sup>2</sup> Email texts shown are the ones administered through a larger online study, for which participants were recruited through an email call to all first-year students at the School of Business and Social Sciences. Instructions were subjects were recruited over the Cognition and Behavior Lab are analogous.

[Link]

Use a desktop computer, notebook or an iPad to participate in this study. Unfortunately, it is not possible to use a smartphone (such as an iPhone or BlackBerry).

Many thanks for participating in this study,  
Alexander Koch (Institut for Økonomi, Aarhus Universitet)

### **Daily emails: Monday – Friday at 0:00, week 2 (Treatment Daily, Daily14)**

Dear [FirstName LastName],

Last week, you agreed to participate in the third part of the scientific study on students' traits, behaviors and study outcomes.

**You set yourself the goal to count [goal Monday] tables today (Monday, [date]).**

To log in and count tables please click on the following link (or copy it into your internet browser):

[Link]

You can use your personalized link to return as often as you like until Friday, [date] - 23:59h. Use a desktop computer, notebook or an iPad to participate in this study. Unfortunately, it is not possible to use a smartphone (such as an iPhone or BlackBerry).

Many thanks for participating in this study,  
Alexander Koch (Institut for Økonomi, Aarhus University)

### **Daily emails: Monday – Friday at 0:00, week 2 (Treatment Weekly)**

As above except:

**You set yourself the goal to count [goal Total] tables until Friday, [date] - 23:59h.**

### **Daily emails: Monday – Friday at 0:00, week 2 (Treatment WeeklyStart)**

*Sorry if you have received this twice. But there have been some problems sending from an au.dk email to certain email addresses such as Hotmail (if you are interested, see all the way below for the message from IT). For that reason, I am sending this again from my gmail account, to be on the safe side.*

Dear [First Name Last Name],

Last week, you agreed to participate in the third part of the scientific study on students' traits, behaviors and study outcomes.

**You set yourself the goal to count [goal Monday] tables today (Monday, [date]).**

**Remember that you need to log on and count at least one table today! \***

To log in and count tables please click on the following link (or copy it into your internet browser):

[Link]

\* The only condition to be eligible to receive any payment (up to 500 kr.) is to log on at least once a day (Monday to Friday) and to correctly count at least one table every day. You get no payment for the first table in a day that you correctly count.

You can use your personalized link to return as often as you like until Friday, [date]- 23:59h. Use a desktop computer, notebook or an iPad to participate in this study. Unfortunately, it is not possible to use a smartphone (such as an iPhone or BlackBerry).

Many thanks for participating in this study,  
Alexander Koch (Institut for Økonomi, Aarhus University)

**Daily emails: Monday – Friday at 0:00, week 2 (Treatment WeeklyStart)**

As above, except:

**You set yourself the goal to count [goal Total] tables until Friday, [date] - 23:59h.**

**Daily emails: Monday – Friday at 0:00, week 2 (Treatment NoGoal)**

As above, except:

[The sentence “You set yourself the goal to count (...)” is deleted]

**Daily emails: Monday – Friday at 0:00, week 2 (Treatment DailyAggregated)**

As for treatment Weekly.