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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) naive individual faces a motivational problem because of present-biased preferences. Using an online, real-effort experiment that varied whether subjects set separate daily goals for how much to work over a one-week period or one weekly goal, we find support for the theoretical predictions. Subjects with daily goals set higher goals in aggregate and provided more effort than those with a weekly goal. The higher effort was driven by the higher goals set. Additional treatments complemented internal commitment through goals with an externally enforced minimum work requirement to get started working each day. Here, average performance dropped because of high dropout.

JEL Classification: D03, D81, D91

Keywords: Self-Control, Goals, narrow bracketing, commitment, Real effort, Online Experiment

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Motivational Goal Bracketing: An Experiment

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1 Introduction

Every day people face motivational problems in repeated tasks such 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. Here a person may focus on single instances of the task and evaluate them 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 often can 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.¹ 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? These open questions are studied theoretically and experimentally in this paper.

We develop a model where a sophisticated or (partially) naïve individual faces a motivational problem because of present-biased preferences, and test its predictions using an online, real-effort experiment. Subjects were randomly assigned to either set 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 for subjects in our experiment, 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.

¹For 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).

We find support for the motivational bracketing hypothesis. Specifically, our model predicts that the aggregated daily goals are higher than the weekly goal. As a consequence, subjects with daily goals should work more than those with the weekly goal. The reason is the following. A weekly goal tempts individuals to put in low effort in the beginning of the week and 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. Taking into account evidence on the distribution of present bias, our model predicts that individuals should adopt a lower weekly goal compared to the aggregated goal level that they would chose with daily goals. The reason is that under plausible assumptions a reduction in the goal level leads to less variation in effort over time. Our data reveal that, indeed, 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 extend our two baseline treatments in two directions. First, we directly test and confirm the prediction that the higher goal level drives the higher effort with daily goals. This is done by comparing effort across 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. This treatment manipulation is of independent interest because it allows us to test predictions of our model that apply also 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. We test and confirm the prediction that subjects with a weekly goal work less in the beginning of the week than subjects with daily goals who face the same aggregated goal level, and that they work more in the end of week to make up for the shortfall.

These findings contribute more broadly to understanding the potential benefits from nudging people to narrowly bracket their goals. For example, employees facing monthly performance targets may naturally adopt a monthly goal bracket for themselves. The effort substitution over time predicted by our model and found in the experiment is consistent with the spikes of effort commonly observed at the end of a month in such settings (e.g. Asch, 1990). The prediction that a narrow goal bracket increases ex-ante utility compared to a broad bracket

– because it induces people to work regularly – is consistent with findings from a field experiment by Cadena et al. (2011). They report that task allocation improved after nudging loan officers to narrowly bracket their performance goals and that loan officers reported higher job satisfaction and lower stress levels.

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 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 how externally enforced work requirements interact with internal commitment through goals. Surprisingly, subjects worked less when they were forced to ‘get started’ working each day (in addition to setting a daily goal) than if they just set daily goals. This result can be explained by a large fraction of subjects who dropped 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 are good at getting people started. The pattern is reversed for the treatments with a weekly goal. For those subjects who did not drop out, effort was significantly higher in the treatment with the requirement than in the one without – as one would expect if the requirement gets people started working. But due to an increase in dropout, the overall effort without the requirement was no different from that with the requirement.

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 affects the level of goals that people set for themselves, the actual effort provided, and the pattern of effort over time in repeated tasks; and how the behavior of sophisticated individuals differs from (partially) naïve ones. Previous work gives conditions under which narrow bracketing is optimal in simultaneous tasks (Koch and Nafziger, 2016) and in two times repeated optimal stopping problems (Hsiaw, 2018).² 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 empirical contribution is to compare behavior for narrowly and broadly bracketed goals in repeated tasks. We are aware of two previous studies that consider the impact of a single, self-set goal on the outcome of an (arguably) repeated task: studying. In their field experiments with students who set a grade goal, van Lent and Souverijn (2017) find a positive effect on grade performance and Clark et al. (2016) no effect but a positive effect of an effort goal. 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 feature very low sample sizes.³

²Fischer 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 have the advantage of providing better incentives because parts of the repayment are shifted to a future period. Both effects are not present with a narrow goal.

³In a 4-week weight loss program, Bandura and Simon (1977) assign 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 who effectively used proximal goals were the ones who lost the most weight, the results must be interpreted with caution because they do not rely on exogenous variation. Bandura and Schunk (1981) run a remedial program with children who have severe math deficits. They assign 10 subjects the narrow goal of completing a certain number of problems in each of seven daily, 30-minute sessions and assign 10

With our experiment we also contribute to the literature on externally enforced commitment devices (for an overview see Bryan et al., 2010). Most closely related are studies that consider effort decisions.⁴ 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. In Kaur et al. (2015), workers can set individual work targets that are then externally enforced by the firm. The novel feature of our experiment is to study whether a flexible, externally enforced minimum work requirement can complement self-enforced goals.

By experimentally studying the interaction between internal and external control, we relate to Bénabou and Tirole (2004). In their theoretical model external control might be good because it commits the individual to provide a certain effort, but bad because it undermines self-reputation. While such self-reputation motives are likely to be absent in our experiment (see footnote 20), our results point to another disadvantage of external control: drop-out.

2 Experimental design and procedures

Our study includes seven treatments, summarized in Table 1. 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 and further questions. A further treatment (*NoManipulation*) does not contribute directly to the research question and is discussed in Appendix G. Experimental instructions are in Appendix J.

Design and treatments. Each treatment consisted of a goal setting and a work part. All parts were conducted online. The experiment involved two parts.

Goal setting part. On a Wednesday at midnight, subjects received an email informing them that they could earn up to 500 Danish kroners (\$ 83) by completing a short online questionnaire before Friday midnight, and performing some online tasks in the following week from

other subjects 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 were able to evaluate their progress toward their goal (cf. Kirschenbaum, 1985, p.494).

⁴Other contexts studied for example are smoking or saving (e.g. Giné et al., 2010; Ashraf et al., 2006).

Table 1: Overview of treatments.

	Treatment ^a	Goals set	Goal feedback frame	Min. work requirement ^c	N	Calendar week/year							
						2013				2014			
						39	40	47	48	49	50	45	48
Sections	Daily	daily	daily	no	78	37	26				15		
2-4	Weekly	weekly	weekly	no	77	35	25				17		
Section	Daily(R) ^b	daily	daily	no	75								75
5.1	Aggregated	daily	weekly	no	75								75
Section	DailyRequirement	daily	daily	yes	47			26	12	9			
5.2	WeeklyRequirement	weekly	weekly	yes	45			27	11	7			
Appendix G	NoManipulation	no goals asked		no	71								71

Notes. Total number of subjects 468. ^aSubjects got daily reminders about goals by email (except *NoManipulation*).

^b*Daily(R)* replicates *Daily*. ^cRequirement to complete at least one table per day.

Monday to Friday. A reminder was sent out on Friday 9 am to those who had not responded. In all treatments, subjects first completed a task based on Abeler et al. (2011).⁵ The task required them to count correctly the number of zeros in a series of tables with thirty cells of zeros and ones. 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, they saw an error message and the table appeared again. A table was not recorded as completed until the correct number was entered.⁶ 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 again would receive DKK 0.5 per completed table. They were informed that they would receive each day an email with a personal link leading to a screen where they could complete the tables. Then they answered some questions that made them think about the benefits of working on the task and their work week ahead.

⁵Subjects in *Daily*, *Weekly*, *DailyRequirement*, and *WeeklyRequirement* were recruited from a previous online study (see heading *Subjects* below). For them, this part took place within that study a few weeks before. These subjects however did receive a reminder about the task.

⁶Subjects were not told that we allowed a margin of error of ± 1 .

Subjects finished 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 knew that we would remind them of their goals next 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, and were reminded that they could use this link anytime up 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 in the presentation of the first two lines above the table to be counted (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 out of the 1,000 tables still could be counted, a reminder about the earnings and 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, which should have made it *ex ante* 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 should have made the leisure alternative tempting.

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.⁷ Subjects could use their own desktop,

⁷Patterns in IP addresses suggest that task outsourcing (e.g. to MTurkers) did not occur (Appendix F).

notebook, or touch-pad. We prevented access with a smart phone (using a software filter). With a smart phone, subjects could perhaps have solved a bit of the task here and there (say, while waiting for the bus), which might not have been 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).

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.⁸ 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 1).

3 Theoretical predictions

Our theoretical predictions are based on a setting where a quasi-hyperbolic discounter (Laibson, 1997) 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). 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 utility flows relative to future utility flows. The exponential discount factor δ is set to one.⁹ The present bias

⁸The 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).

⁹Using a similar real-effort task as ours, Augenblick and Rabin (2018) find that subjects are present-biased and have an estimated daily discounting parameter $\delta \approx 1$.

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 $\beta < 1$ but not immediate costs. So self t prefers effort such that

$$\beta b'(e_t^*) = c'(e_t^*) \quad \text{for all } t = 1, \dots, T. \quad (2)$$

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. This can be either be 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.

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 broad goal G . If the effort differs from its goal by z , the individual experiences a corresponding comparison utility $\mu(z) = z$ for $z < 0$, and $\mu(z) = 0$ for $z \geq 0$.¹⁰ 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 $u_{T+1} = \sum_{t=1}^T [b(e_t) + \min\{0, e_t - g_t\}]$.

The equilibrium concept is that of preferred personal equilibrium (Kőszegi and Rabin, 2006). 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 are assumed to be rational in the sense of perception perfection (O'Donoghue and Rabin 1999; 2001). That is, 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

¹⁰Koch and Nafziger (2016) build on Kőszegi and Rabin (2006) and assume that an effort goal induces reference standards for costs and benefits. Here we define comparison utility over the effort dimension because this corresponds to the frame of the experiment.

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.

3.1 Daily goals (narrow goals)

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

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

Note that (2) and (4) imply $e_{max}(\beta) > e_t^*$. The maximal implementable goal 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 goal cannot fall short of e_t^* , because self t will always choose at least this effort level.

Self 0 picks his daily goals for tasks $t = 1, \dots, T$ to maximize his utility $\beta [b(g_t) - c(g_t)]$ subject to $g_t \in [e_t^*, e_{max}(\hat{\beta})]$. The maximal implementable effort $e_{max}(\beta)$, defined by (4), is increasing in β . A naïve individual overestimates what goals are realistic, because the perceived maximal implementable effort $e_{max}(\hat{\beta})$ exceeds the actual $e_{max}(\beta)$ for $\hat{\beta} > \beta$. Yet, this mistake has no consequences as long as the goal $g_t \leq e_{max}(\beta)$. If the individual sets an overly ambitious goal $g_t > e_{max}(\beta)$, he will underperform relative to that goal. The following result summarizes these insights. All proofs are in Appendix A.

Proposition 1 *Given his belief $\hat{\beta} \geq \beta$, self 0 picks among the implementable narrow goals the one that maximizes his utility: $\max_{g_t \in [e_t^*, e_{max}(\hat{\beta})]} \beta [b(g_t) - c(g_t)]$.*

1. *For β large enough, $e_{max}(\beta) \geq e_0^*$. Self 0 sets daily goals equal to his preferred effort: $g_t^* = e_0^*$ for $t = 1, \dots, T$. The goals are achieved, no matter whether the individual is sophisticated or naïve.*

2. For lower values of β , $e_{max}(\beta) < e_0^*$. Self 0 sets daily goals $g_t^* = \min\{e_0^*, e_{max}(\hat{\beta})\}$ and selves $t = 1, \dots, T$ provide effort $e_t = e_{max}(\beta)$, 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. If the individual is sophisticated ($\hat{\beta} = \beta$), the goal $g_t^* = e_{max}(\beta)$ is achieved. If the individual is (partially) naïve, the goal $g_t^* > e_{max}(\beta)$ and is not achieved.

3.2 Weekly goal (broad goal)

Lemma 1 *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 $\mathcal{A}_t = \emptyset$, $e_t = e_{max}(\beta)$.*

The first part reflects that the incentives to deviate from the goal in a single period are the same under a weekly goal and under daily goals, if all other selves stick to the plan. If self $t \in \{1, \dots, T\}$ 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 $e_{max}(\beta)$. Consequently, a weekly goal cannot improve self-regulation relative to daily goals.

But a weekly goal can harm self-regulation. Unlike with daily goals, self 0 may not be able to implement certain effort profiles because selves $t \in \{1, \dots, T\}$ would deviate, for example, by lowering effort today and compensating with increased effort tomorrow. We refer to such behavior as *effort substitution*. Our next result shows that self 0 can only implement an increasing effort profile when $e_0^* < e_{max}(\beta)$. In particular, self 0 cannot implement his preferred daily effort e_0^* in each period, except in the special case where $e_0^* = e_{max}(\beta)$. In the latter case, each self $t \in \{1, \dots, T - 1\}$ is committed not to lower his effort because future selves will not compensate.

To provide some intuition, suppose the individual is sophisticated ($\hat{\beta} = \beta$), self 0 sets a weekly goal that equals the sum of his desired daily efforts, $G = T e_0^*$, 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 $e_{max}(\beta)$ (Lemma 1). Anticipating that self T will work harder to make-up for any shortfall, it indeed pays off for self $T - 1$ to work less than e_0^* . 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 $G = T e_0^*$. We formalize this intuition in Proposition 4 in Appendix A.3.3. The next result extends the argument to any weekly goal G .

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

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

A naïve individual sets goals in the same way as sophisticated individual would for $\hat{\beta}$. However, the actual effort pattern now differs from the one that self 0 anticipated. First, self $\tau > 0$ applies $\beta < \hat{\beta}$ when deciding on the effort e_τ , whereas self 0 anticipated that $\hat{\beta}$ would be applied. 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 efforts than self 0 anticipated. In that case, self τ will plan to compensate for the short-fall and beliefs will 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. (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: $e_1 = \min\{e_0^*, e_{max}(\beta)\}$.) We formally state this result in 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 no longer is achievable, the individual will not realize this until the point

comes where even $e_{max}(\hat{\beta})$ in every following period would not be enough to achieve the goal. In our parametric example in Figure 1, for example, the individual would only realize in the final period that he will fail the goal.

How does this 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 for example satisfied if the benefits are linear and costs are quadratic.¹¹ With this, we can also compare goal achievement compared to a situation where the individual sets daily goals.

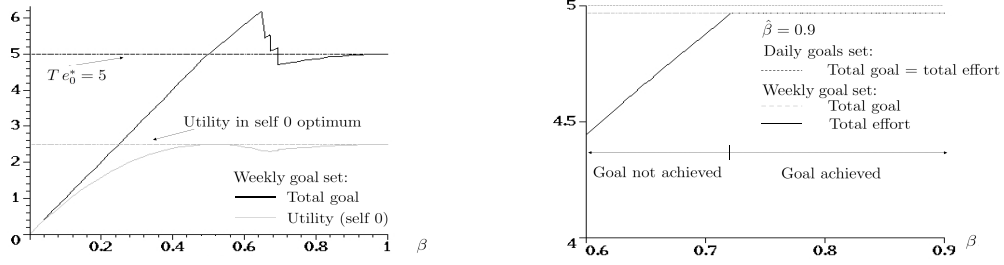
Proposition 3 *Suppose the individual sets a weekly goal, $\hat{\beta} b'''(e) - c'''(e) \geq 0$ and $b'''(e) - c'''(e) \geq 0$. Define $\check{\beta} : e_{max}(\check{\beta}) = e_0^*$.*

1. *There exists $\bar{\beta} \in (\check{\beta}, 1)$, such that for $\hat{\beta} \geq \bar{\beta}$ self 0 sets a total goal $G^* < T e_0^*$ and anticipates interior effort $\hat{e}_{t,0} < e_{max}(\hat{\beta})$ for every period $t = 1, \dots, T$.*
2. *There exists $\underline{\beta} \in (\check{\beta}, \bar{\beta})$, such that for $\hat{\beta} \leq \underline{\beta}$ self 0 sets a total goal $G^* = (T - 1) e_{max}(\hat{\beta}) + e_0^*$ and anticipates a corner solution for periods $t > 1 : (e_0^*, e_{max}(\hat{\beta}), \dots, e_{max}(\hat{\beta}))$.*
3. *A (partially) naïve individual ($1 \geq \hat{\beta} > \beta$) is weakly more likely to fail the weekly goal compared to a situation where he sets daily goals.*
 - (a) *For $e_{max}(\beta) \geq e_0^*$, daily goals would be achieved ($g_t^* = e_0^* = e_t$), but the individual may fail the weekly goal: The individual only achieves the goal if self T faces a remaining part of the goal $G_T = G - \sum_{t=1}^{T-1} e_t \leq e_{max}(\beta)$. This is violated in case of a corner solution with anticipated effort $\hat{e}_{T,0} = e_{max}(\hat{\beta})$ as $e_{max}(\hat{\beta}) > e_{max}(\beta)$. For an interior solution with $\hat{e}_{T,0} < e_{max}(\hat{\beta})$, even though self 0 sets $G^* < T e_0^*$, self T faces $G_T > e_0^*$ which might exceed $e_{max}(\beta)$.*
 - (b) *For $e_{max}(\beta) < e_0^*$, the individual fails to reach his goals under both goal setting formats.*

In principle, self 0 could 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

¹¹Augenblick and Rabin (2018) estimate effort costs for a similar real-effort task as ours, finding an approximately quadratic cost function.

Figure 1: Total goal, utility of self 0, and total effort (T=5).



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

Table 2: Hypotheses and summary of findings

	Hypothesis	Finding	Section	Table
H1	$Total\ goal(Daily) > total\ goal(Weekly)$	✓	4.1	4
H2	$Total\ effort(Daily) > total\ effort(Weekly)$	✓	4.2	4
H3a	$Goal\ achievement\ rate(Daily) = goal\ achievement\ rate(Weekly)$	✓	4.2	A3
	$Goal\ achievement\ rate(Daily(R)) = goal\ achievement\ rate(Aggregated)$	✓	5.1	A3
H3b	Controlling for $total\ goal$, $total\ effort(Daily) = total\ effort(Weekly)$	✓	4.2	4
	$Total\ effort(Daily(R)) = total\ effort(Aggregated)$ (as $total\ goal$ constant)	✓	5.1	A10
H4	Goal non-achievers closer to total goal in <i>Daily</i> than in <i>Weekly</i>	✓	4.2	A4
	Goal non-achievers closer to total goal in <i>Daily(R)</i> than in <i>Aggregated</i>	✗	5.1	A4
H5a	$Monday\ effort(Daily) > Monday\ effort(Weekly)$	✗	4.3	A11
	$Monday\ effort(Daily(R)) > Monday\ effort(Aggregated)$	✓	5.1	A12
H5b	$Friday\ effort(Daily) < Friday\ effort(Weekly)$	✗	4.3	A12
	$Friday\ effort(Daily(R)) < Friday\ effort(Aggregated)$	✓	5.1	A12

Note: Tables prefaced with 'A' are in the appendix.

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 he is sufficiently optimistic about his present bias. 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 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 β 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.

3.3 Parametric example

We illustrate our results with a parametric example with $b(e) = e$ and $c(e) = \frac{1}{2}e^2$. Here, $e_0^* = 1$, $e_1^* = \beta$, and $e_{max} = 2\beta$. With daily goals, self 0 can implement his desired effort e_0^* if $\beta \geq \frac{1}{2}$. The left panel of Figure 1 shows how self 0 would set the weekly goal G^* as a function of $\hat{\beta}$ if $T = 5$, like in our experiment (see Appendix B for details).

Augenblick et al. (2015) estimate a population present bias parameter of $\beta = 0.9$ using a similar real-effort task as ours, and find that almost all mass of present-biased individuals lies on $\beta \in [0.6, 1)$ (see their Figure VI).¹² Augenblick and Rabin (2018) estimate $\beta = 0.83$ and $\hat{\beta} \approx 1$. Individual estimates for present-biased individuals are concentrated on $\beta \in [0.5, 1)$ and $\hat{\beta}$ are concentrated around 1 (see their Figure 6). They find only a weak correlation between β and $\hat{\beta}$.

In the right panel of Figure 1, we therefore consider a partially naïve individual with $\hat{\beta} = 0.9$ and the actual effort response for $\beta \in [0.6, 1]$. Notice that the goal is achieved except in the case where the individual severely overestimates β . But in any case, the total goal and total effort would be lower than if the individual set daily goals, where $\sum_t^T g_t^* = \sum_t^T e_t = 5$.

If a lot of individuals had $\hat{\beta}$ close to $\check{\beta}$, where $e_{max}(\check{\beta}) = e_0^*$, the weekly goal G^* could exceed the total daily goal $T e_0^*$ (Part 2 of Proposition 3, Figure 1). Yet, the evidence on β 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 most relevant case, and that few subjects are likely to so severely overestimate β to drive a wedge between goal achievement in *Weekly* vs. *Daily* (Part 3a of Proposition 3). With these observations, we reach the hypotheses stated in Table 2.

¹²Some 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 β 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.

4 Daily goals vs. weekly goal

Data. We have data on 468 participants. Our focus here are the 155 subjects in the main treatments *Daily* and *Weekly* (Table 1). Primary outcome variables are *total goal* (the aggregated daily goals in *Daily*, the weekly goal in *Weekly*), *total effort* (the total number of correctly counted tables), *effort* on a given weekday, and *goal achievement* (a dummy variable whether $total\ effort \geq total\ goal$). Secondary outcome variables, such as log-ins and task completion time allow us to study some mechanisms. Control variables available for all treatments are a *gender* dummy and *baseline productivity* (balance tests are in Table A2). 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 highly collinear with the goals that subjects set and therefore not used as control variables. For *Daily*, *Weekly*, *DailyRequirement*, *WeeklyRequirement* 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 E.

Censoring. Our design aimed to make high effort desirable from self 0 perspective. In line with this, roughly half of the subjects in *Daily* and *Weekly* chose a goal of completing all 1,000 tables and 30 to 41 percent of subjects completed all 1,000 tables (Table 3). A common way of dealing with such censoring is to employ tobit regressions, which we do in the following. Further robustness checks in Appendix I indicate that results are robust to relaxing assumptions of the tobit model.

4.1 Goals

In line with Hypothesis H1, subjects set higher goals in *Daily* than in *Weekly* (Table 3). Subjects in *Daily* aimed to complete a total of 789 tables on average, whereas those in *Weekly* only aimed for 682 (Fisher-Pitman permutation test for two independent samples, $p = 0.041$).¹³ In tobit regressions, this difference is significant only after controlling for

¹³A non-parametric test of difference in distributions, henceforth simply referred to as permutation test. We ran Monte Carlo simulations with 10,000 repetitions, using Kaiser’s (2007) implementation `permtest2`.

Table 3: Descriptive statistics.

Treatment	N	Average total		Fraction of subjects with			Average		
		goal	effort	goal =1000	effort =1000	effort <goal	effort -goal	number of logins ^a	effort per login
Daily	78	789	690	0.51	0.41	0.45	-98.54	6.72	131.45
Weekly	77	682	521	0.47	0.30	0.47	-161.03	5.52	104.06
Daily(R)	75	796	572	0.41	0.43	0.53	-224.12	8.16	97.97
Aggregated	75	791	649	0.48	0.43	0.39	-141.88	6.32	99.26
DailyRequirement	47	858	487	0.53	0.36	0.60	-370.45	6.32	83.60
WeeklyRequirement	45	750	558	0.53	0.42	0.40	-191.62	7.60	76.06

Notes. Effort: tables correctly counted. ^a New login: table solved on a new day or >30 min. since last entry.

gender and productivity – indicating that the coefficients are imprecisely estimated without the controls (Table 4). Looking at the distribution of goals shown in the left panel of Figure 2, *Daily* had more mass on high goals than *Weekly*. For example, 65 percent aimed for at least 800 tables in the former compared to 48 percent in the latter.

4.2 Effort

Subjects took on average 15 seconds to complete a table and 90 percent of subjects required 20 seconds or less on average per table.¹⁴ More productive subjects were slightly faster (OLS coefficient -0.390, se 0.124, $p = 0.002$) with a one standard deviation higher baseline productivity cutting completion time by 1.4 seconds. Subjects made 37 mistakes ± 1 away from the correct solution on average. Subjects were not told that we allowed this error margin. For technical reasons, we could not record mistakes that required recounting of a table.

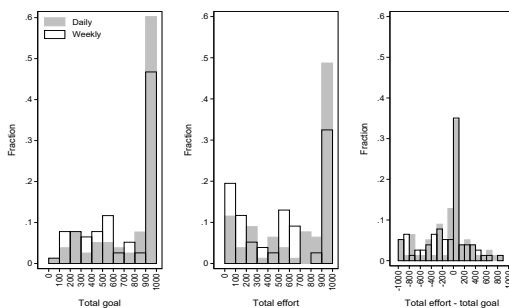
In line with Hypothesis H2, subjects provided more effort in *Daily* than in *Weekly* on average: 690 vs. 521 tables (Table 3; permutation test, $p = 0.005$). This difference is significant also in tobit regressions (Table 4) and is also reflected in the distribution of effort. The middle panel of Figure 2 shows that *Daily* has more mass on high effort than *Weekly*.

The effort gap between treatments could be driven by differences in goal setting and differ-

Permutation tests and Fisher’s exact tests reported below are all against the two-sided null hypothesis of no difference.

¹⁴We truncate completion time at 120 seconds to take out effects of breaks/stopping to log on again later.

Figure 2: Distribution of goals and effort (*Daily* & *Weekly*).



Notes. Bins are half-open intervals, except for the last one: $[0, 100)$, $[100, 200)$, \dots , $[900, 1000]$.

ences in goal achievement. Or perhaps daily goals on their own were more motivating than a weekly goal. For example, because daily goals are more timely and weekly goals are more distant (Bandura and Simon 1977; see also Appendix D).

Differences in total goal. Once we control for the total goal, the treatment difference in effort becomes smaller and is no longer consistently significant (Table 4). 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 H3b. Another indication that *Daily* increased effort by exogenously shifting goals upwards relative to *Weekly*, is that the instrumental variables 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, $p = 0.039$).¹⁵

Goal achievement. According to our model, naïve individuals overestimate how much effort they will be able to put in at the end of the week, since $e_{max}(\hat{\beta}) > e_{max}(\beta)$. However, in many cases this should not lead to a treatment difference in goal achievement. First, if $e_0^* > e_{max}(\beta)$, the individual would set a too high total goal and fail it in both treatments. Consistent with this, goal achievers had a lower total goal than non-achievers (permutation test, $p = 0.089$, but not robustly significant in tobit regressions available on request). Second, if $e_0^* \leq e_{max}(\beta)$, the individual would achieve his total goal in *Daily*. For a sufficiently low β and a sufficiently biased belief $\hat{\beta}$, the individual would engage in excessive effort substitution and fail his goal in *Weekly*, as illustrated in Figure 1. Yet, such combinations of β and $\hat{\beta}$

¹⁵We thank an anonymous referee for suggesting this.

seem unlikely given the evidence discussed in Section 3.3. Thus, we expect only a negligible treatment difference. Consistent with Hypothesis H3a, we find no treatment difference in goal achievement (Fisher’s exact test, $p = 0.872$; Table A3).

In addition, goal non-achievers are predicted to be closer to their total goal in *Daily* than in *Weekly*, because effort substitution causes the individual to put in low effort at the start of the week in *Weekly* whereas in *Daily* each self t will go to the maximal implementable effort $e_{max}(\beta)$. In line with Hypothesis H4, the right panel of Figure 2 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 A4).¹⁶ Around 29 percent hit within ± 1 of their total goal (28 percent in *Daily* and 30 percent in *Weekly*). For those with *total goal* < 1000, the numbers are 9 percent overall, 3 percent in *Daily*, and 15 percent in *Weekly*.

4.3 Effort substitution

Figure 3 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 H5a,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 $e_T = e_{max,T}(\beta)$ is more likely than with constant effort costs (see Appendix C).

The data from *Daily* reveal some heterogeneity in goal profiles. 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

¹⁶Because, $e_{max}(\beta) > 0$ for $\beta > 0$, the argument behind Hypothesis H4 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: 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 ^a	150.95 (96.82)	173.14* (89.37)	186.89** (92.25)	235.46** (96.97)	228.39** (94.86)	180.91*** (42.41)	151.44* (91.31)	100.89 (92.44)
Total goal							0.60*** (0.14)	0.60*** (0.15)
Baseline productivity		24.17** (11.08)	10.88 (19.00)		33.52*** (10.67)	67.04*** (10.52)	24.49** (10.38)	57.82** (22.41)
Female		-417.57*** (90.65)	-448.27*** (102.71)		-93.66 (99.86)	-176.73*** (52.36)	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 (287.80)	-230.05 (168.23)	-349.58 (616.96)
Full controls ^b	no	no	yes	no	no	yes	no	yes
Margin.effect(daily goals) ^c	75.14	89.44* (90.65)	97.36** (102.71)	135.79** (71.05)	134.38** (176.97)	110.07*** (287.80)	93.93* (168.23)	64.15 (616.96)
Effect size ^d	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.

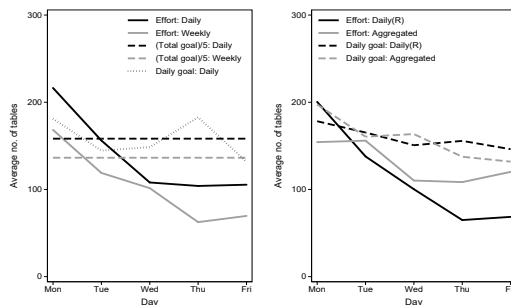
* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bSee Appendix E. ^cTobit marginal effect on the censored latent variable (at the means of control variables). ^d $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly}}$.

Table 5: Transition from goal profiles to effort profiles in *Daily*.

Goal profiles	Effort profiles			
	Flat ^d	High-low ^e	Low-high ^f	Other
Flat ^a	5.88	70.59	17.65	5.88
High-low ^b	0	77.42	16.13	6.45
Low-high ^c	0	48.15	48.15	3.70
Other	0	66.67	33.33	0

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

Figure 3: Average daily effort and goals.



average goal for the last two days), and 35 percent aimed to ‘start low and end high’ (Table A5). 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 5). The picture is similar for goal achievers and non-achievers (Table A7). This suggests that some subjects failed to predict 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 A8).

The left panel of Figure 3 suggests that subjects in *Weekly* worked less on Monday than subjects in *Daily*, as predicted by Hypothesis H5a. 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 H5b (Tables A11 and A12). 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 above, 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. To address the issue in a clean way, we ran further treatments *Aggregated* and *Daily(R)*, discussed in the next section. These allow to control for the daily goals that subjects initially set, and here we find the predicted patterns of effort substitution (right panel of Figure 3; Section 5.1). We summarize our findings in Table 2.

5 Extensions

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

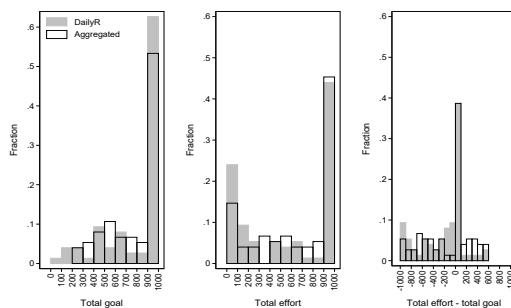
By comparing *Daily* and *Weekly* we could examine the overall effect of the goal format on goals and effort. Yet, as the goal level was endogenous, we can only partly conclude (by controlling for *total goal* in the regressions) whether the difference in effort was caused by the goal bracket (daily vs. weekly) or by the difference in goals that the different treatments induced (Hypothesis H3b). 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 H5a,b).

To address these issues, we conducted treatments *Daily(R)* and *Aggregated* that (i) allow us to compare effort across treatments ‘holding fixed’ the total goal and (ii) provide us with daily goals for both treatments that we can use as controls to obtain a cleaner test of effort substitution. The treatments are of independent interest also, because showing the effect of exogenously shifting the framing (but not the level) of goals is highly relevant to organizations (see our discussion of the field experiment by Cadena et al. (2011) in Section 7).

Treatments. All subjects set daily goals like in *Daily*. Then they were randomized either to *Daily(R)* (where they 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*.

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 A14). 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

Figure 4: Distribution of goals and effort (*Daily(R)* & *Aggregated*).



Notes. Bins are half-open intervals, except for the last one: $[0, 100), [100, 200), \dots, [900, 1000]$.

not with the feedback format weekly goal (*Aggregated*). This is what we find (Table A9). In our model, self 0 sets as daily goal $g_t = e_0^*$, unless the present bias is so severe that self 0 thinks only $e_{max}(\hat{\beta}) < e_0^*$ is implementable. Given the evidence on present bias discussed in Section 3.3, we would expect a typical individual to face the aggregate goal $G = T e_0^*$ after being randomized into *Aggregated*. It is straightforward from the proofs of Propositions 2 and 5 that Hypotheses H3a, H4, and H5a,b continue to apply, whereas H3b should be true because the total goal should not vary across *Daily(R)* and *Aggregated* by design.

Total effort. Our first objective was to revisit Hypothesis H3b. Because the treatment did not affect the total goal, there should be no treatment effect on total effort – which is what we find (permutation test, $p = 0.247$; Table A10).

Goal achievement. In line with Hypothesis H3a, 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 A3). In line with H4, the right panel of Figure 4 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 those subjects who fell short of the total goal but attempted working ($total\ effort > 0$) reveal no significant treatment effect (Table A4).

Effort substitution. Our second objective was to cleanly test for effort substitution, by exploiting daily goals as a control for possible heterogeneity in effort costs over the course of the week. In line with Hypothesis H5a, the right panel of Figure 3 suggests that subjects in

Aggregated worked less on Monday than subjects in *Daily*. While a permutation test has $p = 0.154$, the tobit coefficient becomes significant once adding controls (Table A11). 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 H5b. This is confirmed by a permutation test ($p = 0.017$) and tobit regressions (Table A12). Our treatments offer an additional way to test Hypothesis H5b. 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, $p \leq 0.020$).

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 (‘getting started’) and then continue to work towards their goal (‘getting 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, Mann-Whitney test, $p = 0.0997$)¹⁷ and subjects in *Daily* log in on more weekdays than subjects in *Weekly* (3.2 vs. 2.6, Mann-Whitney test, $p = 0.010$). Once logged on, subjects with daily goals completed 131 tables per login in *Daily* vs. 104 in *Weekly* (permutation test, $p = 0.106$).

To explore further whether getting started helps subjects to get finished, we conducted treatments *DailyRequirement* and *WeeklyRequirement*. They differ 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 if or 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

¹⁷Note that the Mann-Whitney test compares the distributions and not the averages.

earnings for the week.¹⁸

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 (‘get finished’). 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 (Table 3; permutation test, $p = 0.067$). The treatment difference is not significant in tobit regressions though. Further, we observe no significant difference in effort (permutation test, $p = 0.448$; Table A13), and there is no treatment difference in the number of logins (Mann-Whitney test, $p = 0.436$), weekdays logged on (Mann-Whitney test, $p = 0.806$), or tables completed per login (permutation test, $p = 0.717$). The results suggest the interpretation that the motivational power of a daily goal comes from helping people to get started working regularly. If the work requirement does the same, there should be no treatment difference in goals or effort for *DailyRequirement* compared to *WeeklyRequirement*. That the minimum work requirement indeed helps to get started is supported by the fact that completing only one table is very rare (14 single-table-for-a-day logins out of 460 subject-day combinations). But our analysis in the next subsection qualifies this positive message.

5.2.2 Do work requirements complement internal commitment?

The comparisons of *Daily* and *DailyRequirement*, as well as *Weekly* and *WeeklyRequirement*, allow us to examine whether an externally enforced minimum work requirement complements internal commitment through goals.

¹⁸The 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).

Predictions (without a formal theory). Ariely and Wertenbroch (2002) outline that externally imposed commitment harms time-consistent individuals because it takes away flexibility. In our setting, subjects only had to spend less than 1 minute per day (turn the computer on, click the emailed link, solve one table). The work requirement still allowed to flexibly distribute effort over the five workdays. Only those subjects who anticipated to be away from a computer with internet access for a whole day – a situation that is quite unlikely during term time – should have dropped out. On the other hand, Ariely and Wertenbroch (2002) outline that externally imposed commitment can benefit time-inconsistent individuals. If subjects thought that the work requirement would help them to get started or work more regularly, they should have set a higher total goal and then provided 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). 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 do some exploratory analysis of the dropout behavior in the different treatments. Our hypothesis was that the work requirement did not take away much flexibility and thus we expected similar dropout rates in the treatments with and without the requirement. The experimental data contradict this. Table 7 shows that the dropout probability increased from around 4 percent (*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. Further, there was no significant effect of the requirement on the likelihood of working on Monday (bottom of Table 7). Thus, rather than incentivizing subjects to get started earlier, it appears to have pushed those subjects to drop out who otherwise would have started working later than Monday.

In a next step, we examine whether the work requirement had any benefits for those 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, $p =$

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

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

Notes. Test of H_0 : minimum work requirement did not affect the total goal/total effort, holding fixed the goal setting format. 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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aAll: *Daily*, *Weekly*, *DailyRequirement*, *WeeklyRequirement*. No dropout: subsample with total effort > 0/satisfying daily login requirement. ^bDummy for *DailyRequirement* or *WeeklyRequirement*, respectively. ^cTobit marginal effect on the censored latent variable of the minimum work requirement vs. no requirement (at the means of control variables). ^d $\frac{\text{Margin.effect(requirement)}}{\text{Standard deviation of outcome in } \textit{Daily} \text{ or } \textit{Weekly}, \text{ respectively}}$.

Table 7: Dropout

Treatment	N	Drop out/zero effort ^a		Login Monday		Total effort (mean)	
		N	Percent	N	Percent	Full sample	Not dropped out
Daily	78	3	3.8	62	79.5	690.0	717.6
Weekly	77	3	3.9	51	66.2	520.8	541.9
DailyRequirement	47	17	36.2	35	74.5	487.1	744.9
WeeklyRequirement	45	15	33.3	32	71.1	558.3	811.1
Marginal effect of minimum work requirement vs. no requirement (percentage points). ^b							
		Drop out/zero effort ^a		Login Monday			
DailyRequirement vs. Daily ^c		30.13*** (7.53)		-5.71 (8.12)			
WeeklyRequirement vs. Weekly ^c		29.10*** (7.58)		4.26 (8.67)			

Notes. ^aDropout: counted zero tables or, if relevant, failed to satisfy the daily login requirement. ^bEach 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 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*, $N = 247$. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^cDummy for *DailyRequirement* or *WeeklyRequirement*, respectively.

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*.¹⁹ 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 those individuals who do not drop out, helping them to ‘get started’.

6 Discussion

Goal non-achievement. Almost half of the subjects failed to reach their goal (Table 3). This seems 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 1). 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. However, those who set such a ‘subjectively realistic goal’ were not more likely to achieve their goal (Fisher’s exact test, $p = 0.145$; logit marginal effect 9.44 percentage points, $p = 0.134$, $N = 305$). Even if it did, still only 58 percent of these subjects achieved their goal.

The descriptive evidence in Section 4.3 suggests that many subjects may have failed to correctly predict their behavior or changes in circumstances later in the week. Irrespective

¹⁹Mon: 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*.

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 5). 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 4.3). Theoretically, this would have a similar effect as naïveté about the present bias. One can easily see this in our parametric example with quadratic effort costs, where c_t has an inversely proportional effect to β on $e_{max,t}(\beta) = 2\beta/c_t$. 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: from 75 to 59 percent in *Daily* and from 51 to 37 percent in *Weekly* (for those who had not yet reached their total goal before Friday).

Mistakes in goal setting could explain our findings if they led subjects to set a higher goal in aggregate 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 $p > 0.445$). 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*, which we can proxy for using a question that subjects answered right before learning that they should set goals. They estimated the maximum number of tables they would be able to complete if they used all the spare time they had available to work on the tasks. 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 $p > 0.913$).

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 1000th table

compared to the first table). To test for possible effects of experience on goal setting and achievement, we compare 71 out of 78 subjects in *Daily*, 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, $p > 0.242$).

Wrong beliefs or uncertainty about effort costs could explain some of the observed goal non-achievement (next to, or in combination with partial naïvité). Consider first the possibility of wrong beliefs. 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 $\beta b'(\bar{e}_{max}) + \beta = \bar{c}'(\bar{e}_{max})$). 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.

Now consider uncertainty about effort costs. In a theoretical model, Koch and Nafziger (2016) demonstrate the effect of uncertainty on narrowly or broadly bracketed goals. A broad bracket allows to pool 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*.

Next, consider learning about effort costs occurring gradually over time. 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. Revelation of uncertainty or learning about effort costs might lead subjects to revise their goals. 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, the prediction of effort substitution remains also when allowing for goal revision. So 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).

Power. We performed power calculations only ex-post. Figure A4 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 power of a two-sided test was 0.54, 0.77, or 0.85, respectively, for a significance level of 0.01, 0.05, or 0.1. That is, depending on the significance level, there was a 54 to 85 percent chance that we would fail to reject the null hypothesis of no treatment difference even if, in fact, the effort in *Daily* is lower than that in *Weekly*. Thus our main study was underpowered for $\alpha \leq 0.05$, while for $\alpha = 0.1$ it was correctly powered. For *Daily* vs. *Aggregated* we have a similar sample size. Comparisons involving treatments *WeeklyRequirement* and *DailyRequirement* however have lower sample sizes and hence are underpowered for conventional significance levels. On account of the limited power of tests, some care should be taken when interpreting statistically insignificant findings that however show non-negligible effect sizes (reported in the tables). Specifically, we cannot rule out that the coefficients are imprecisely estimated and that subjects set higher goals in *DailyRequirement* than in *WeeklyRequirement* (Table 3) or that that subjects set higher goals when facing the minimum work requirement (Table 6) – in all these cases there are non-negligible effect sizes around 0.2.

Gender effects. Table 4 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., 2016). 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 (Mann-Whitney test, $p < 0.001$, $N = 155$), less likely than men to say about themselves that they set “ambitious goals” ($p = 0.096$), and more often than men said that they avoided setting goals because they were afraid not to achieve them ($p = 0.005$). A systematic exploration of gender differences in goal setting and achievement appears interesting for future research.

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.²⁰ From a prior study (Epper et al., 2018) we have information whether a subject would set a daily, weekly, overall, or no goal in an exam preparation vignette. A dummy variable for the answer ‘no goal’ (as opposed to some kind of goal) is correlated with dropout in the treatments with the work requirement *DailyRequirement* and *WeeklyRequirement* (logit marginal effect 45-56 percentage points, $p < 0.001$, $N = 92$), but not robustly significant for the treatments without the work requirement *Daily* and *Weekly* (10-26 percentage points, $p = 0.087$ without controls, $p = 0.572$ with full controls, $N = 155$). This suggests that dropout might have been a response to having been forced to work every day and not so much a response to having been forced to set goals. This is consistent with the literature which tends to find that externally imposed goals increase performance (see, among others, Goerg and Kube, 2012). In particular, subjects might have reacted to having been *forced* to count *one* table. It is obvious that completing this one table is not an ambitious target. Therefore it might have been perceived as a nuisance rather than as an encouragement. Other external requirements or incentive schemes might have worked better. For example, one could have imposed a monetary fine on the subjects if they did

²⁰Another explanation builds on Bénabou and Tirole (2004), where external control is a bad signal about effort costs and crowds out intrinsic motivation. We have no measure available to test this explanation. However, we believe that intrinsic motivation for the counting task is likely to have been low from the start.

not fulfil a certain workload, thus moving away from a minimum effort rule to an externally enforced non-trivial goal. How to best design such requirements is an interesting avenue for future research.

Length of the experiment. The length 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 β 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 students studying for an exam. With longer time horizons, goals may be less salient and the issue of reminders may be relevant. An advantage of daily goals could be that they act as reminders by themselves and that they help the individuals to organize their schedules. 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 this requires keeping an account for accumulated effort.

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

As discussed above, one reason for goal non-achievement might have been that subjects revised their goals. An interesting treatment for future research would be to explicitly give subjects the opportunity to revise their goals. This might open up for interesting

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 add here by controlling the type of uncertainty that subjects face, by studying the exact effort patterns (for daily vs. weekly goals), or by studying whether goal revision can explain goal non-achievement.

A further interesting 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, etc. 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* would however 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. It thus seems likely that they would perform better than subjects in *Weekly* – also because they would be able to respond to deviations from their previous daily goals caused by naïveté about their present bias.

Another potential effect of goal bracketing might be that achieving a weekly goal is cognitively more challenging than achieving a daily goal. After all, 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. For example, one could just divide the weekly goal by the number of days, or one could infer the likely profile of daily goals from other subjects. But our study revealed a great deal of heterogeneity in goal profiles (Section 4.3). The problem is that the way one disaggregates the weekly goal is likely to affect the results.

7 Conclusion

We provide a theoretical framework and experimental evidence for the motivational benefits of narrowly bracketed, 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 (narrow) goals than under a weekly (broad) goal. The increase in effort was primarily related to the higher level of goals set when goals were bracketed narrowly rather than broadly. In additional treatments we exogenously shifted the framing (but not the level) of goals. 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.

Many organizations struggle with a suboptimal task allocation of their employees over time, often manifested in spikes of effort immediately prior to a bonus deadline (e.g. Asch, 1990). Our theoretical and empirical results suggest that reframing a bonus threshold in terms of smaller and more frequent narrow goals can have beneficial effects. This is supported by evidence from a field experiment at a Columbian bank by Cadena et al. (2011). 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 effort allocation over time was suboptimal. It caused a mismatch in the timing of cash flows that increased costs of cash flow management for the bank and, according to personnel surveys, it contributed to nearly 70 percent of loan officers feeling stressed at work, mostly during the second half of the month. As in our experiment, more narrowly bracketed goals led to better effort allocation over time without affecting the overall output. 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.²¹ 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

²¹The intervention offered the chance to win small non-cash rewards by completing credit placements during the first two weeks of each month. These rewards were independent of the monthly bonus structure and represented only 2 percent of the average compensation. Importantly, results were only significant in the second half of the intervention during which branch managers additionally reminded loan officers of their weekly goals and progress toward the bonus targets.

loan officers' portfolios. In addition, workers reported higher job satisfaction and lower stress levels. The intervention was only temporary and the effects disappeared once the feedback about weekly goals was discontinued.

Our findings tie in, more generally, with evidence on motivational benefits from externally inducing subjects to narrowly bracket 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 envelop. In the treatment condition, the wage payments were split into two envelopes, with one containing the amount that the worker expressed as a savings goal. Laborers in the treatment condition saved more than those in the baseline. Soman and Gourville (2001) find that self-selected single-performance ticket holders, who likely narrowly focus on seeing a single theatre performance, were more likely to carry through with this ex ante desirable choice 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. The demand for such commitment could explain the puzzle why people prefer microcredit over microsavings.

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Online Appendix for “Motivational Goal Bracketing: An Experiment”

A Proofs

A.1 Proof of Proposition 1

The maximal implementable goals $e_{max}(\beta)$ and $e_{max}(\hat{\beta})$ are defined by (4) by plugging in β 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 naïf 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 Proof of Lemma 1

Consider a weekly goal G . The utility of self τ for an effort level e_τ and anticipated effort responses $\hat{e}_{t,\tau}$ for $t > \tau$ that overall lead to failing G is

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

Taking the derivative w.r.t. e_τ , (5) is increasing for $e_\tau \leq e_{max,\tau}^b(\beta)$, 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 τ will provide up to $e_{max,\tau}^b(\beta)$. We now show that $e_{max,\tau}^b(\beta) = e_{max}(\beta)$ for all τ .

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 τ 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 sophisticated individuals facing a weekly goal

We start by proving some intermediate results on goal achievement and effort patterns for sophisticated ($\hat{\beta} = \beta$) individuals that will be used in the proofs of Propositions 2, 3, and 5. For partially naïve individuals ($\hat{\beta} > \beta$), 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 efforts and therefore coincides with (2). That is, he would chose $e_t = e_t^*$. But as $T e_t^* \leq G$ 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 τ 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.

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 ϵ shortfall in effort. Note that for ϵ 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 ϵ 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 works more than is optimal from the perspective of self 0 and self 1 works less: $e_T > e_0^*$ 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 follows in part from Lemma 6. Below we 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 $e_T > e_0^*$.²²

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

²²For 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. Part 3 follows directly from the explanations given in the proposition and from Proposition 5 (stated in Appendix A.6).

Step 3: Implemented effort profile

Lemma 10

(i) For β 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 β 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 β .

(ii) $U_0(e_0^*, e_{max}(\beta), \dots, e_{max}(\beta))$ is continuous in β 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 β 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 (Partially) naïve individuals 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 sophisticated individuals ($\hat{\beta} = \beta$) pins down the beliefs of self 0 about future effort.

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

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

where $\hat{e}_{t,\tau}$ denotes the belief of self τ about the effort of self $t > \tau$. Specifically, self τ 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 τ 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 τ concludes, after observing previous efforts, 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 $e_5 = \min\{G_5 - \sum_{t=1}^4 e_t, 2\beta\}$, $e_4 = \begin{cases} \min\{\frac{\beta}{\mathbf{a}} G_4, 2\beta\} & \text{if } G_4 - e_4 < 2\beta, \\ \min\{G_4 - 2\beta, 2\beta\} & \text{if } \hat{e}_{5,4} = 2\beta, \end{cases}$

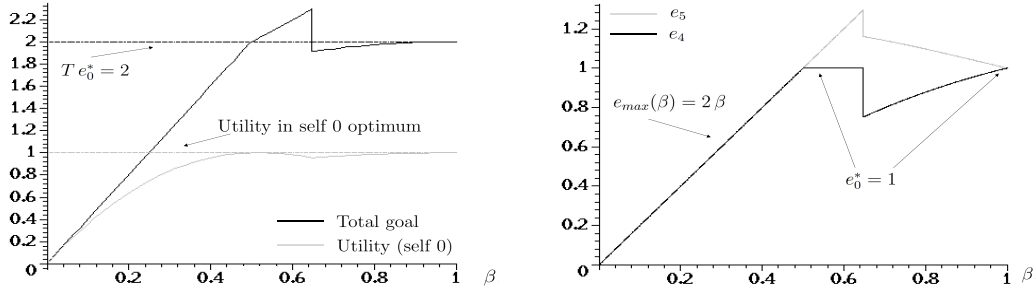
$$\begin{aligned} e_3 &= \begin{cases} \min\{\frac{\mathbf{b}}{\mathbf{B}} G_3, 2\beta\} & \text{if } G_3 - \hat{e}_{4,3} - e_3 < 2\beta \text{ and } \hat{e}_{4,3} < 2\beta, \\ \min\{\frac{\beta}{\mathbf{a}}(G_3 - 2\beta), 2\beta\} & \text{if } G_3 - 2\beta - e_3 < 2\beta \text{ and } \hat{e}_{5,3} = 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_2 - 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 ($T=2$).



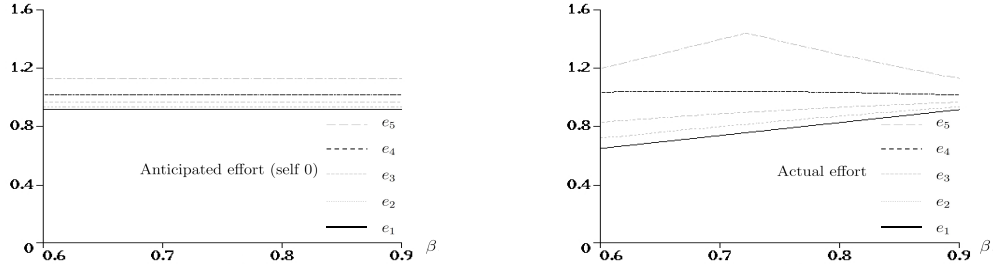
obtain the utility of self 0 as a function of β . Note that for a given β not all cases may be relevant. For example, for $\beta < 0.5$, case (v) is not feasible because $e_0^* = 1 < e_{max}(\beta) = 2\beta$. Further, note that the utility (and as illustrated below, the optimal goal and the associated effort profiles) are not monotone in β . 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 ϵ , 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 β , 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., β 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 β but dominates the interior solution for β less than (approximately) 0.65.

B.2 Partially naïve individual

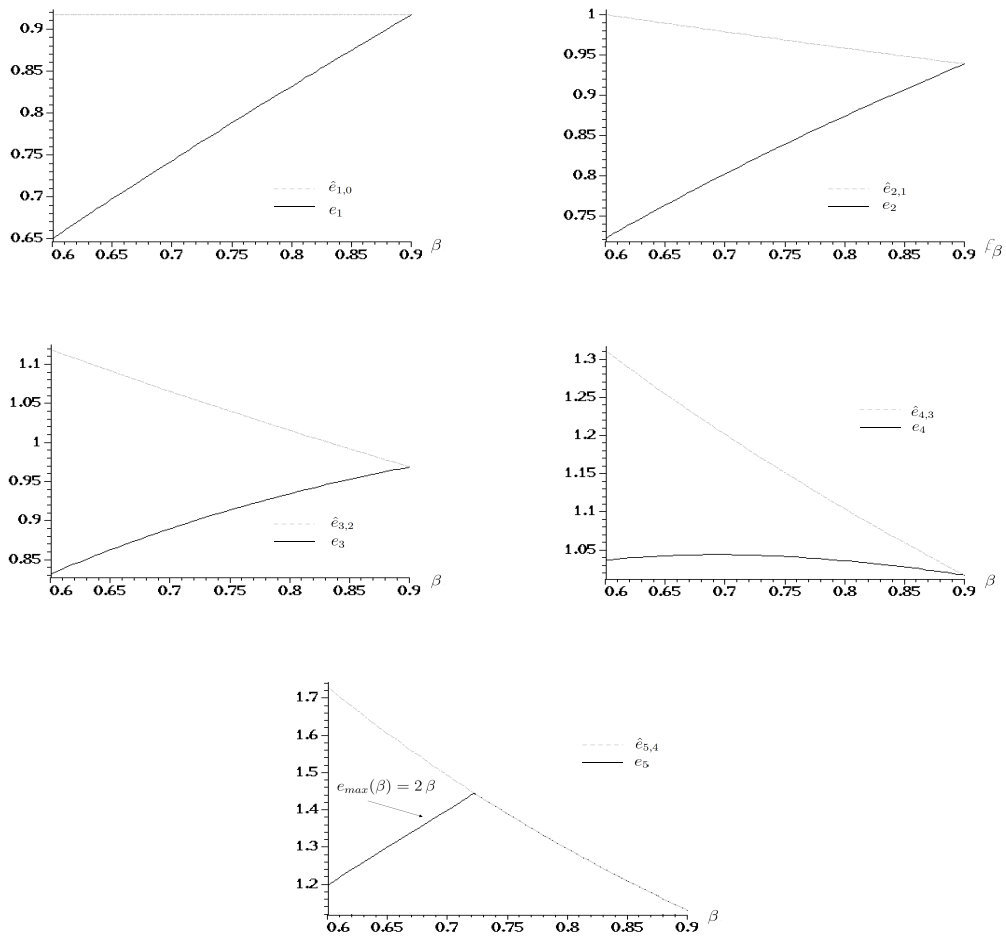
Just as in the case of a sophisticated individual, effort is determined by recursively substituting in. The difference is that self t applies the actual β to current period calculations but applies $\hat{\beta}$ when determining anticipated effort $\hat{e}_{\tau,t}$ for $\tau > t$. A partially naïve 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 1 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 3) 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 efforts $\hat{e}_{t,0} \leq e_{max,t}(\hat{\beta})$ 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 β , 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 Additional control variables

Subjects for treatments *Daily*, *Weekly*, *WeeklyRequirement* and *Daily Requirement* 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. τ : scaling parameter for error 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 Joliffe (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.

F 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.²³ 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.²⁴ This could have been due to subjects logging in, e.g. from their parents. 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, $p = 0.32$).

G 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.

²³One 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.

²⁴A 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.

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*.²⁵

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.

²⁵From 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 ^a	63.25 (104.89)	29.04 (102.13)	-16.56 (136.07)	-9.42 (129.93)
Weekly goal ^b	-108.15 (112.10)	-148.22 (108.59)		
Controls ^c	no	yes	no	yes
Margin.effect(daily goals) ^d	33.09	22.90	-7.68	9.58
$\frac{\text{Margin.effect(daily goals)}}{\text{std.dev.NoManipulation(total effort)}}$	0.08	0.06	-0.02	0.02
Margin.effect(weekly goal) ^d	-56.58	-67.06		
$\frac{\text{Margin.effect(weekly goal)}}{\text{std.dev.NoManipulation(total effort)}}$	-0.14	-0.17		
N	301	300	146	145

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

H Supplementary tables

Table A2: Test of balance across treatment pairs.

Treatments	Daily		Daily(R)	DailyCommitment	
	(vs. Weekly)		(vs. Aggregated)	(vs. WeeklyCommitment)	
Variables ^a	(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
τ		-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). * $p < 0.10$,

** $p < 0.05$, *** $p < 0.01$. ^a See Appendix E for explanations of controls in (2) & (5): .

Table A3: 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 ^a	0.02 (0.08)	0.01 (0.08)	-0.05 (0.10)	0.04 (0.08)	-0.01 (0.10)			
Daily goals feedback ^b						-0.15* (0.08)	-0.10 (0.09)	-0.10 (0.09)
Total goal				-0.0003* (0.0001)	-0.0003* (0.0002)			0.0002 (0.0002)
Baseline productivity		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 ^c	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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cFull controls: see Appendix E.

Table A4: 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 ^a	112.98 (72.29)	126.60* (65.86)	134.77** (63.56)		
Daily goals feedback ^b				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 ^b	no	no	yes	no	no
Margin.effect(daily goals) ^c	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 ($0 < \text{total effort} < \text{total goal}$). Tobit coefficient (marginal effect on the latent dependent variable) with robust standard error in parenthesis. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cSee Appendix E. ^dTobit marginal effect on the censored latent variable (at the means of control variables).

Table A5: Profiles of goals (*Daily*).

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

$g_{mon} \leq g_{tue} \leq g_{wed} \leq g_{thu} \leq g_{fri}$, at least one inequality strict. ^eExcludes 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 A6: Profiles of goals (all treatments with daily goals).

Profile	N	Percent
‘Flat’ ($g_{mon} = g_{tue} = \dots = g_{fri}$)	47	20.61
Daily goal = 200	26	11.40
Daily goal = 100	9	3.95
Other	12	5.26
‘High-low’ ($(g_{mon} + g_{tue})/2 > (g_{thu} + g_{fri})/2$)	100	43.86
$(g_{mon} + g_{tue})/2 \geq g_{wed} \geq (g_{thu} + g_{fri})/2$ ^a	45	19.74
Decreasing ^b	19	8.33
One daily goal = 1000 ^c	2	0.88
‘Low-high’ ($(g_{mon} + g_{tue})/2 < (g_{thu} + g_{fri})/2$)	71	31.14
$(g_{mon} + g_{tue})/2 \leq g_{wed} \leq (g_{thu} + g_{fri})/2$ ^a	30	13.16
Increasing ^d	10	4.39
Other type of profile	10	4.39
All	228	100
All daily goals > 0	195	85.53
At least one daily goal = 0 ^e	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*. ^aAt least one inequality strict. ^b $g_{mon} \geq g_{tue} \geq g_{wed} \geq g_{thu} \geq g_{fri}$, at least one inequality strict. ^cIn both cases, $g_{mon} = 1000$. ^d $g_{mon} \leq g_{tue} \leq g_{wed} \leq g_{thu} \leq g_{fri}$, at least one inequality strict. ^e 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 A7: Transition from goal profiles to effort profiles in *Daily*.

Goal profiles		Effort profiles			
		Flat ^d	High-low ^e	Low-high ^f	Other
Goal	Flat ^a	11.11	55.56	33.33	0
achievers ^g	High-low ^b	0	85.71	14.29	0
	Low-high ^c	0	50.00	50.00	0
	Other	0	50.00	50.00	0
Goal	Flat ^a	0	87.50	0	12.50
non-achievers ^g	High-low ^b	0	70.59	17.65	11.76
	Low-high ^c	0	44.44	44.44	11.11
	Other	0	100	0	0

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

Table A8: 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
‘Flat’ ($e_{mon} = e_{tue} = \dots = e_{fri}$)	1	2.33	0	0.00	0	0.00	0	0.00
Daily effort = 200	1	2.33	0	0.00	0	0.00	0	0.00
‘High-low’ ($(e_{mon} + e_{tue})/2 > (e_{thu} + e_{fri})/2$)	27	62.79	24	68.57	25	60.98	21	58.33
$(e_{mon} + e_{tue})/2 \geq e_{wed} \geq (e_{thu} + e_{fri})/2$ ^a	12	27.91	14	40.00	13	31.71	12	33.33
Decreasing ^b	7	16.28	11	31.43	5	12.20	7	19.44
One daily effort = 1000 ^c	4	9.30	0	0.00	3	7.32	0	0.00
‘Low-high’ ($(e_{mon} + e_{tue})/2 < (e_{thu} + e_{fri})/2$)	15	34.88	7	20.00	14	34.15	11	30.56
$(e_{mon} + e_{tue})/2 \leq e_{wed} \leq (e_{thu} + e_{fri})/2$ ^a	4	9.30	2	5.71	5	12.20	5	13.89
Increasing ^d	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
Total effort = 0	0	0.00	3	8.57	0	0.00	3	8.33
Other type of profile	0	0.00	1	2.86	2	4.88	1	2.78
All	43	100	35	100	41	100	36	100
All daily efforts > 0	15	34.88	9	25.71	4	9.76	0	0.00
At least one daily effort = 0 ^e	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. ^aAt least one inequality strict. ^b $e_{mon} \geq e_{tue} \geq e_{wed} \geq e_{thu} \geq e_{fri}$, at least one inequality strict. ^cIn all cases, $e_{mon} = 1000$. ^d $e_{mon} \leq e_{tue} \leq e_{wed} \leq e_{thu} \leq e_{fri}$, at least one inequality strict. ^eExcludes 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 A9: 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 ^a	(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 ^a	(0.16)	(0.29)	(0.28)	(0.36)	(0.34)
N	150	149 ^b	139 ^b	133 ^b	120 ^b

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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aInteraction between treatment dummy and the daily goal level. ^bExcludes subjects whose cumulative completed tables reached 1000 on the previous day.

Table A10: Impact of goal feedback format on total effort (*Daily(R)* vs. *Aggregated*).

	(1)	(2)	(3)
Daily goals feedback ^a	-108.26	-17.45	-37.37
	(132.05)	(130.91)	(122.85)
Total goal			1.07***
			(0.19)
Baseline productivity		51.72***	38.69**
		(16.73)	(16.03)
Female		247.14*	299.74**
		(132.17)	(124.93)
Constant	856.74***	-166.93	-823.58***
	(96.52)	(311.69)	(309.98)
Margin.effect(daily goals feedback) ^b	-50.22	-8.40	-19.09
Effect size ^c	-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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily(R)*. ^bTobit marginal effect on the censored latent variable (at the means of control variables). ^c $\frac{\text{Margin.effect(daily goals feedback)}}{\text{Standard deviation of total goal in Aggregated}}$.

Table A11: 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) ^f	(9)
Daily goals ^a	85.48 (56.14)	93.47* (56.10)	102.05* (57.75)	65.99 (56.10)	73.29 (56.54)				
Daily goals feedback ^b						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) ^d	56.01	61.62*	68.63*	43.90	49.81	45.20	64.88**	63.75**	64.40**
Effect size ^e	0.24	0.24	0.29	0.19	0.21	0.27	0.38	0.38	0.38
Controls ^c	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).

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cbaseline productivity and a gender

dummy; full controls: see Appendix E. ^dTobit marginal effect on the censored latent variable (at the means of control

variables). ^e $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of effort on Monday in } Weekly \text{ or } Aggregated, \text{ respectively}}$. ^fReplacing *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.

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

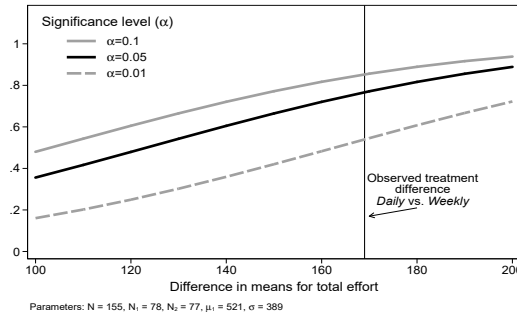


Table A12: Impact of goal and goal setting or feedback format on effort on Friday.

Treatments	Daily vs. Weekly					Daily(R) vs. Aggregated			
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8) ^f	(9)
Daily goals ^a	110.48*	102.60*	127.03**	96.51*	122.57**				
	(56.09)	(55.52)	(56.68)	(56.84)	(57.21)				
Daily goals feedback ^b						-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) ^d	44.65	41.23*	48.95*	38.81	47.22	-58.71	-48.12**	-50.50**	-53.93**
Effect size ^e	0.30	0.30	0.32	0.26	0.31	-0.41	-0.33	-0.35	-0.37
Controls ^c	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).

* $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cbaseline productivity and a gender

dummy; full controls: see Appendix E. ^dTobit marginal effect on the censored latent variable (at the means of control vari-

ables). ^e $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of effort on Friday in Weekly or Aggregated, respectively}}$. ^fReplacing *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 A13: 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)	(6)
Daily goals ^a	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) ^b	69.59	-35.54	-62.62	29.17	20.94	
Effect size ^c	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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$.

^aDummy for *DailyRequirement*. ^bTobit marginal effect on the censored latent variable (at the means of control variables).

^c $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of outcome in } WeeklyRequirement}$

I 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 A14 -A17 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 A14: Impact of goal setting format or goal feedback format on 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 ^a	118.19** (48.45)	173.14* (89.37)	193.40 (146.92)	399.84** (189.39)				
Daily goals feedback ^b					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) ^c	118.19*	89.44**			7.90	-8.41		
Effect size ^d	0.35	0.27			0.03	-0.03		
N	155	155	155(105) ^e	155(88) ^e	150	150	150(117) ^e	150(85) ^e

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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cOLS coefficient/tobit marginal effect on the censored latent variable (at the means of control variables). ^d $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in } Weekly \text{ or } Aggregated, \text{ respectively}}$. ^eNumber of observations effectively used in the CLAD/SCLS estimation in parenthesis.

Table A15: 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 ^a	164.03*** (60.35)	228.39** (94.86)	239.08 (131.49)	638.10*** (126.45)				
Daily goals feedback ^b					-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) ^c	164.03***	134.38***			-37.92	-8.40		
Effect size ^d	0.42	0.34			-0.10	-0.02		
N	155	155	155(150) ^e	155(88) ^e	150	150	150(144) ^e	150(134) ^e

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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cOLS coefficient/tobit marginal effect on the censored latent variable (at the means of control variables). ^d $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly or Aggregated, respectively}}$. ^eNumber of observations effectively used in the CLAD/SCLS estimation in parenthesis.

Table A16: 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	Tobit	CLAD	SCLS	OLS	Tobit	CLAD	SCLS
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Daily goals ^a	111.14*	151.44*	187.08**	201.96				
	(58.18)	(91.31)	(99.73)	(124.89)				
Daily goals feedback ^b					-42.01	-37.37	-0.00	-66.60
					(62.18)	(122.85)	(127.21)	(132.80)
Total goal	0.45***	0.60***	0.82***	0.68***	0.52***	1.07***	0.89***	0.84***
	(0.09)	(0.14)	(0.16)	(0.17)	(0.10)	(0.19)	(0.13)	(0.21)
Productivity	15.01**	24.49**	18.60*	18.92	16.89**	38.69**	0.00	29.15**
	(6.17)	(10.38)	(12.09)	(11.99)	(7.29)	(16.03)	(19.31)	(14.59)
Female	73.22	53.56	40.46	48.54	144.71**	299.74**	249.00	147.26
	(61.24)	(99.11)	(108.76)	(121.68)	(65.13)	(124.93)	(163.20)	(136.01)
Constant	-68.01	-230.05	-319.90	-208.67***	-136.65	-823.58***	-141.77	-426.54***
	(109.22)	(168.23)	(189.61)	(30.84)	(140.22)	(309.98)	(321.17)	(90.94)
Margin.effect(daily goals) ^c	111.14*	93.93*			-42.01	-19.09		
Effect size ^d	0.28	0.24			-0.11	-0.05		
N	155	155	155(131) ^e	155(129) ^e	150	150	150(150) ^e	150(124) ^e

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. * $p < 0.10$, ** $p < 0.05$, *** $p < 0.01$. ^aDummy for *Daily*. ^bDummy for *Daily(R)*. ^cOLS coefficient/tobit marginal effect on the censored latent variable (at the means of control variables). ^d $\frac{\text{Margin.effect(daily goals)}}{\text{Standard deviation of total goal in Weekly or Aggregated, respectively}}$. ^eNumber of observations effectively used in the CLAD/SCLS estimation in parenthesis.

Table A17: Getting started: Effort on Monday (OLS regressions).

Treatments	Daily vs. Weekly				DailyRequirement WeeklyRequirement				Daily(R) vs. Aggregated	
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Daily goals ^a	51.71 (39.53)	54.78 (42.80)	30.96 (40.18)	32.15 (42.46)	29.90 (41.94)	64.21 (50.29)	16.77 (40.99)	42.44 (49.25)		
Daily goals feedback ^b									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 productivity	6.77 (5.95)	21.12* (12.66)	4.50 (5.69)	18.78 (12.27)	2.88 (4.71)	3.74 (11.48)	2.67 (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)	-20.58 (47.89)	-60.69 (63.56)	-16.70 (46.98)	-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)	132.46 (84.07)	-21.46 (385.84)	42.54 (82.18)	-175.85 (397.96)	-111.18* (66.05)	-236.36*** (64.78)
Full controls ^c	no	yes	no	yes	no	yes	no	yes	no	no
Effect size ^d	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	91	78	91	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). $p < 0.10$, $**p < 0.05$, $***p < 0.01$. ^aDummy for *Daily* or *DailyRequirement*. ^bDummy for *Daily(R)*. ^cFull controls: see Appendix E.

^dCoefficient(daily goals)
Standard deviation of effort on Monday in Treatment *Weekly*, *WeeklyRequirement*, *Aggregated* or *Weekly*, respectively.

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J Instructions¹

[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

¹ 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) 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.**

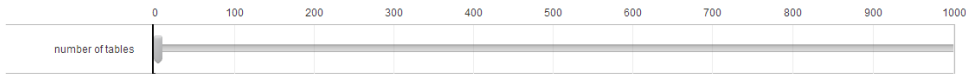
On the next page you will set yourself a goal for how much you want to work next week.

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?



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:

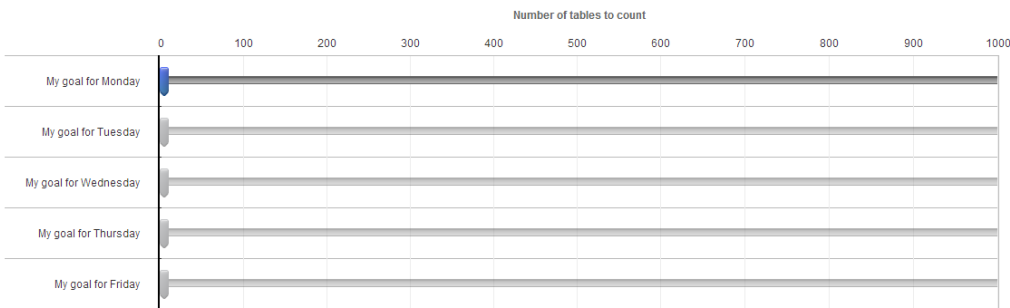
Screen 3 (Treatment Daily, Daily14, DailyAggregated)

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.

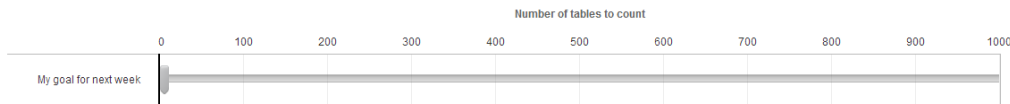


Screen 3 (Treatment Weekly)

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 texts²

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]

[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]):

² 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.