## Field and Online Experiments on Procrastination and Willpower<sup>\*</sup>

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#### Abstract

Self-control problems have recently received considerable attention from economic theorists. We conducted two studies to test the benefits of externally imposed deadlines and how willpower depletion affects behavior, providing some of the first data in these areas. Each study involved a behavioral intervention designed to affect performance. We find that for a lengthy task, regular deadlines neither reduce procrastination nor increase completion rates. Second, a willpower-depleting task reduces initial effort but increases overall task-completion rates. Our results help to inform ongoing efforts to understand and model procrastination, willpower and commitment mechanisms.

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### 1. INTRODUCTION

The issues of self-control, procrastination, commitment, and willpower have important economic consequences in realms such as productivity, savings behavior, and well-being, and these have been the focus of a number of recent theoretical models. People experience self-control problems when their preferences are not consistent across time. One form of self-control problem concerns persistent bad habits or addictions, such as overeating or cigarette smoking. An individual knows that he or she will later regret a current self-indulgent choice, but nevertheless engages in the activity. The other side of the coin is a situation in which an individual is faced with an activity that will lead to future benefits, but is unappealing at the moment. This often leads to procrastination, common in everyday life.<sup>1</sup> People vow to stop smoking, stop eating ice cream, or start exercising... tomorrow. An important issue is the underlying nature of the self-control problem and how one can overcome it; closely related to this issue is the notion of willpower.

In this paper, we examine aspects of commitment and willpower in two experimental studies featuring behavioral interventions. As the duration of the tasks made it rather infeasible to use simple laboratory experiments, we instead conducted field and online experiments. In Study 1, we investigate how people allocate their time over a task of significant duration (up to five weeks) and whether an externally-imposed commitment in the form of binding sub-goals would help them to complete the overall task. I

Having observed possible willpower effects in Study 1, we chose to focus on willpower in Study 2. Here we consider the effects of willpower depletion on performance; in one treatment, we implemented a mechanism to deplete willpower—a Stroop test—on the first day of

<sup>&</sup>lt;sup>1</sup> Procrastination has been found to be quite pervasive among students: Ellis and Knaus (1977) find that 95 percent of college students procrastinate, while Solomon and Rothblum (1984) find that 46 percent nearly always or always procrastinate in writing a term paper.

a task that could be spread over two days.<sup>2</sup> We expected willpower depletion to have a negative effect on performance.

Extant empirical research on procrastination and self-control is significant (e.g., DellaVigna and Malmendier 2004 and Ashraf, Karlan, and Yin 2006 in the economics literature), but experimental economics has contributed less to this body of work. Given that experiments offer a way to test interventions in controlled settings, they may offer the best approach to exploring these behavioral phenomena in detail. As there is little or no experimental evidence in the economics literature on patterns of effort provision for tasks of significant duration, our results represent some of the first detailed and daily data regarding procrastination and willpower under controlled conditions and with financial incentives and tasks of significant duration and elapsed time.<sup>3</sup>. One primary purpose of our research is to identify patterns in behavior that will inform theorists so that more relevant, descriptive models can be developed; at the same time, we also hope to aid other researchers in designing mechanisms that are effective in overcoming obstacles to performance and to explore how experimental methods can shed light on a set of robust but competing theories.<sup>4</sup>

There have been only a few studies that consider how one might *overcome* self-control problems. Aside from exerting willpower in the face of a disagreeable task, one approach is commitment, or binding one's own behavior with costly restrictions. Wertenbroch (1998)

<sup>&</sup>lt;sup>2</sup> Psychologists have used the Stroop test (Stroop, 1935) to deplete willpower for many years (see Gailliot *et al.* 2007 for a recent study and MacLeod 1991 for a history).

<sup>&</sup>lt;sup>3</sup> Two recent studies consider issues of temptation. Houser, Reiley, and Urbancic (2007) look at temptation at the check-out counter in the supermarket, while Houser, Schunk, Winter, and Xiao (2009) conduct a laboratory experiment involving temptation and commitment.

<sup>&</sup>lt;sup>4</sup> There have been at least a handful of behavioral interventions designed to overcome bad habits or to form new good ones. Angrist and Lavy (2009) offer substantial cash incentives in Israel for matriculation; while this is ineffective when individual students are selected for the treatment, matriculation rates do increase when this program is school-wide. Charness and Gneezy (2009) pay students at two American universities to attend a gym during a period of time, finding that attendance rates increase substantially not only during this period, but also after the intervention ends. Angrist, Lang, and Oreopoulos (2009) offer merit scholarships to undergraduates at a Canadian university, with some success in improving performance, but mixed results overall.

presents examples of binding behavior, including tactics such as putting savings into a Christmas-club account that does not pay interest or buying only small packages of goods such as cigarettes or ice cream. In a similar vein, Burger and Lynham (2009) examine weight-loss bets in England, where one could bet on achieving a weight goal by a deadline; however, the vast majority of bettors lost their bets with the agency, suggesting that this self-imposed deadline was ineffective.

Ariely and Wertenbroch (2002) consider the effectiveness of deadlines in tasks that will almost certainly be completed.<sup>5</sup> They present two studies in which three tasks had to be completed over a period of time, finding that externally-imposed costly deadlines during this period are more effective than self-imposed (and binding) costly deadlines, which in turn are more effective than having no additional deadlines. This suggests that exogenous deadlines might well be quite effective; however, the commitment level in these experiments was not extreme, in that the cost for not meeting an interim deadline was rather low. Moreover, the cost of not completing the "study"—dropping out of class—was quite high.

The results of our behavioral intervention in Study 1 differ considerably from the Ariely and Wertenbroch (2002) results, suggesting that some deadlines may in fact reduce flexibility enough to overcome any benefits they provide in reducing procrastination. Completion rates in our first study were actually 50 percent *higher* with no interim deadlines, implying that our form of externally-imposed commitment was ineffective.<sup>6</sup> The patterns of effort show a pronounced weekly cycle, with little difference in behavior from the first week to the fifth week. Overall,

<sup>&</sup>lt;sup>5</sup> The tasks involved turning in term papers for grades in a class for a 14-week period (with a one percent grade penalty for each day of delay) or proofreading three papers totaling about 100 pages over a 3-week period; in the latter task, participants were paid \$0.10 to find errors (there were about 100 embedded), but fined \$1.00 for each day of delay.

<sup>&</sup>lt;sup>6</sup> Of course, one might argue that our weekly requirements were not well chosen. However, we chose these after considerable discussion with undergraduate students and *ex post* would choose them again.

there is no aggregate support in our data for strongly present-biased preferences, although behavior seems consistent with a combination of moderate present-biased preferences and uncertainty over the cost (or feasibility) of future effort. In any case, individual analysis reveals substantial heterogeneity, with some people exerting most of their effort early on and other people concentrating their effort into the last weeks.

In our second study, we offered participants a financial reward for completing a task within two days. On the first day, subjects were assigned either a willpower-depleting or a willpower-neutral Stroop test. Among the people who completed the task, those who were assigned the willpower-depleting test exerted significantly less effort on the first day, as expected. However, those subjects who were assigned the willpower-depleting test were actually *more* likely to complete the overall task within the allotted two days than the people in the willpower-neutral treatment. Thus, while willpower does appear to be depleted by the intervention on the first day, it appears to come back even stronger on the second day; this is consistent with the notion that successfully completing a willpower-depleting Stroop test serves as a signal about one's willpower.<sup>7</sup> We discuss this finding in more detail later. Again, we see considerable heterogeneity across the population.

There are a number of theoretical approaches that address procrastination, timeinconsistency, and willpower; these include models of present-biased quasi-hyperbolic preferences (Laibson, 1997 and O'Donoghue and Rabin, 1999), temptation and commitment (Gul and Pesendorfer, 2001 and Fudenberg and Levine, 2006), self-signaling of one's strength of will (Bénabou and Tirole, 2004), and willpower (Ozdenoren, Salant, and Silverman, forthcoming). While these models do relate to our experiments, it is not our purpose to conduct

<sup>&</sup>lt;sup>7</sup> An alternative explanation is that participants who completed the Stroop test viewed this as a form of commitment.

<sup>&</sup>lt;sup>9</sup> We thank ORSEE (Greiner 2004) for the free recruiting software, which permitted selective invitations.

a systematic test of these theories. Nevertheless, we do highlight where our results are either broadly consistent with or contradictory in spirit to the relevant theoretical work. Because the predictions of these models depend on (unknown) parameter values, we also administered a questionnaire in order to calibrate some of the parameters relating to the quasi-hyperbolic model. This questionnaire focuses on aspects of the tasks in both of our experiments. In the remainder of this paper, we provide details of our experimental design in section 2 and describe the results in section 3. We offer some discussion in section 4, and conclude in section 5.

### 2. THE EXPERIMENTS

### Study 1

This experiment was conducted at the University of California at Santa Barbara. We obtained permission to have anonymous access to the grade records of students in a large introductory undergraduate class and then recruited as many as possible from this class. We then advertised the session to first-year students in the general experimental subject pool.<sup>9</sup> All students were told that they could attend an introductory meeting about an experiment that would involve a non-trivial amount of money to be earned over time. Interested students were randomly assigned to one of two introductory meetings.<sup>10</sup> Participation was voluntary and everyone who showed up was guaranteed \$5 even if they were not interested in participating in the study. At these meetings, we explained the nature and rules of the experiment. This process lead to a total of 74 eventual participants (out of 87 students who showed up to the meetings); 42 were from the class and 32 were from the campus-wide experimental subject pool. As we show later, there was no appreciable difference in behavior across these two sets of participants.

<sup>&</sup>lt;sup>10</sup> All the students in a particular informational meeting were assigned to the same treatment group. This was done to reduce social interaction threats (Cook and Campbell, 1979).

We chose the task of studying because it is a common activity for students, but one that is susceptible to procrastination. Studying has obvious longer-term benefits, but is costly in the short-run insofar as other activities have more immediate appeal.<sup>11</sup> Nevertheless, there are already incentives in place for the studier; thus, we did not pay the usual average per-hour rates for experiments, but chose to pay \$95 for 75 hours of monitored studying. Ideally one could record 'natural' studying completed by students, but this was not feasible. Instead, we recorded the time students spent studying in a popular, easy-to-access location: the central campus library.

We showed participants the studying location, a room in the library that was frequently (but intermittently) monitored. This study area was available for between 14 and 16.5 hours each day. Subject to the availability constraint, students were free to log in and out by handing over an ID card to the monitor who would then log the student in or out on a computer. In addition, students were each given a large numbered placard, unique to each individual. This was visible to the monitor at all times. The studying area was monitored hourly at a varying time each hour to ensure students were present at the studying location when signed in.

We conducted two treatments. In one treatment, there were no requirements other than logging in the 75 hours of study over the five-week period. In the other treatment, each student was required to log in a cumulative total of at least 12 hours per week (12 hours by the end of the first week, 24 hours by the end of the second week, etc.); the idea was that this would prevent participants from falling too far behind, while still offering some scope for procrastination (one could log 48 hours by the end of the fourth week, but would still need to log in 27 hours in the last week). If these requirements were not met, one did not receive any payment. Note that this 'forfeiture' penalty is quite different from the 'soft' penalties in Ariely and Wertenbroch (2002).

<sup>&</sup>lt;sup>11</sup> Students may therefore wish to do more studying than they actually manage; this is similar to self-control problems such as dieting or smoking.

Each student was assigned a web page where he or she could check on the number of hours logged, and could then contact us in the case of any discrepancy. In addition, students who satisfied weekly studying requirements 'banked' their contingent earnings; their web pages had a check-like graphic showing the credit already amassed (of course, this credit was only to be paid if the student completed the overall 75-hour requirement). Students who failed to meet a weekly requirement were notified at the end of the applicable week that they were no longer eligible to earn the \$95. At the end of the five-week period, those students who had completed the requirement(s) received their earnings and filled out a short questionnaire.

We would like to immediately address two possible concerns. First, students had access to both computers and wireless Internet, so we cannot be certain how much of their time in the library was devoted to studying. However, the anecdotal evidence from the monitors is that, although students occasionally just checked e-mail or Facebook, etc., apparent studying was by far the most common activity observed. Moreover, our results indicate some improvement in grades among those students who completed the studying task, so there is also an inference that significant studying occurred. Second, one might also be concerned about contamination, since people from both treatments studied in the same area. Again, we have only anecdotal evidence against this: 1) Exit interviews of the people who completed the study task indicate that students who spent over 75 hours in each other's company weren't aware of the other treatment group, and 2) Monitors did not observe students in conversation with one another.

### Study 2

Where our first study collected detailed data on long-term procrastination in a quasirealistic setting, our second study focuses more closely on procrastination over a shorter time period; Study 2 also adds the element of willpower depletion, as some of the patterns observed in

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Study 1 were suggestive of willpower effects. This approach allows us to define procrastination more clearly and examine whether the days of the week for the study affects outcomes.

Study 2 was conducted online with multiple-choice economics questions.<sup>12</sup> We recruited participants from a micro principles class and two intermediate micro classes at UCSB; we supplemented the 135 participants from these classes with 23 economics or business economics majors who were not in these classes, but who were in the campus-wide subject pool. In all cases, we described the task (in general terms) and the payment scheme. The task consisted of answering, over a two-day period, 20 multiple-choice questions taken from microeconomics exams. The participant could answer all 20 of these on the first day or spread them over the two days.<sup>13</sup> We paid each person \$7.50 for completing the Stroop exercise and the 20 multiple-choice questions, with each correct answer earning an additional \$0.75 for the student.

Each participant was randomly assigned to either a Tuesday-Wednesday group or a Friday-Saturday group; these different two-day periods presumably reflect either different opportunity costs or different stocks and/or flows of willpower. We also required people to do 250 rounds of a Stroop test on the first day before proceeding to answer the multiple-choice questions.<sup>14</sup> Discordant Stroop tests are used in psychology experiments to deplete willpower and consist of showing words that are the names of colors, although the actual words are printed in a color of ink different from the color name they represent. For example, the word "blue" might be printed in red ink. One is asked to respond by typing the color seen and ignoring the word itself. Resisting the urge to read the word and instead focus on the color of the ink requires

<sup>&</sup>lt;sup>12</sup> We also conducted a study involving tasks of varying length to be completed online over seven consecutive days. However, the overall completion rate was only 15% and there was evidence of substantial confusion on the part of the participants. We observe smaller day-of-the-week effects and a slight tendency to delay the longer tasks until later. Details of this study are reported in Burger, Charness, and Lynham (2008).

<sup>&</sup>lt;sup>13</sup> It was made clear to each participant that being inactive for 15 minutes after logging on would result in one being logged off for that day; thus, people could not return to the task on the same day after taking a long break.

<sup>&</sup>lt;sup>14</sup> To control for possible differences in the time spent on the two Stroop tasks, we required an elapsed time of 15 minutes to complete the concordant Stroop; this was roughly the time it took to complete the discordant Stroop.

considerable use of the brain's executive functions. We randomly assigned each participant to process either discordant Stroop exercises or concordant ones (where the word color matches the ink color and the task is trivial). Thus, one treatment group had its willpower depleted before beginning work on the multiple-choice questions and the other group did not.

### Questionnaire

The predictions of some of the leading theoretical models within the context of our experiments depend on parameter values. Since these cannot be divined from our experimental results, we conducted an anonymous survey of 69 undergraduate students in an economics class in an attempt to gain some insight into some of the actual parameter values. The full questionnaire is presented in Appendix D. While there was no way to make these questions incentive compatible, we did pay each student \$2 for completing the questionnaire, as anecdotal evidence suggests that such payment induces more serious thinking about the questions and responses.

The questions focus on aspects of the tasks in both of our studies. One question concerns how people would allocate 75 hours of library study time over each week of a 5-week period. In one version, people were asked how they would <u>prefer</u> to allocate their time during this period, while another version asked how they thought they would <u>end up</u> allocating their time during this period. The first version can be used as a benchmark against which procrastination can be measured; if people would ideally have a sharply-increasing profile over time, the presence of such a profile would not indicate that they are in fact procrastinating. Comparing the second version to the first provides some evidence of whether students are aware of any tendency to delay relative to their optimal path, *i.e.* are they "sophisticated" or "naïve" about their tendency to procrastinate? Other questions asked students how they would allocate 15 hours of library

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study over the days of the week, how likely it was they would not be able to complete a project that required periodic work over a 5-week period, how much they would have to be compensated for various amounts of studying, and whether they would rather do 90 minutes of homework all at once or do more homework spread out over two days.

### 3. EXPERIMENTAL RESULTS

### Study 1

One of our goals was to test whether providing interim targets or requirements would help people to achieve the overall task. In this regard, Figure 1 shows the proportion of students who completed the studying project:



Figure 1 - Success rates, by treatment

In the weekly-requirements treatment, 15 of 38 students (39.5%) were successful. When there were no weekly requirements, 22 of 36 students (61.1%) completed the mandated 75 hours. Clearly, there is no support for the view that externally-imposed restrictions helped students to achieve the goal. In fact, the test of the difference of proportions (see Glasnapp and Poggio 1985) gives Z = 1.86, so that there is a marginally-significant difference in success rates across treatments (p = 0.062, two-tailed test).<sup>15</sup>

<sup>&</sup>lt;sup>15</sup> Recall that some participants were from an introductory class and others were first-year students from the general subject pool. Thirteen of 20 students from the class succeeded in the no-weekly treatment, compared to 9 of 16

We also find that the completion rate for the 45 female participants was more than 50 percent higher than for the 29 male participants.<sup>16</sup> However, this difference is only marginally significant; the test of the difference of proportions gives Z = 1.67 (p = 0.095 with a two-tailed test). This result is similar in flavor to the finding in Angrist, Lang, and Oreopoulos (2009) that providing incentives for good grades is much more effective for female students.

One might speculate as to what the completion rate for the students with no weekly requirements would have been if they had faced the requirements. While it is impossible to construct the appropriate counterfactual, by simply imposing *ex post* the weekly requirements on the no-requirements treatment, we find that the completion rates are quite similar. The completion rate for the no-requirements treatment would now be 14 of 36 (38.9 percent), compared to 15 of 38 (39.5 percent) for the weekly-requirements treatment. In fact, the nearly identical 'counterfactual' rates suggest the difference in success rates is driven by the presence of the restrictions and the corresponding reduction in flexibility, implying that procrastination *per se* was unaffected by imposing weekly restrictions.<sup>17</sup>

One interesting aspect of our data is the cyclical weekly patterns in the number of study hours logged. Figure 2 shows these patterns for those students who completed the project (referred to as "winners"). While it may not be surprising to see a weekly, cyclical pattern when students face weekly study requirements, the pattern is actually slightly stronger for the group

students from the subject pool (Z = 0.54), while eight of 22 students from the class succeeded in the weeklyrequirements treatment, compared to 7 of 16 students from the subject pool (Z = 0.46). Neither difference (nor the overall comparison of 21 of 42 versus 16 of 32) is close to statistical significance.

<sup>&</sup>lt;sup>16</sup> Twenty-six of the 45 females (57.8%) completed the requirement(s), compared to 11 of the 29 males (37.9%).

<sup>&</sup>lt;sup>17</sup> On the other hand, a portion of the difference in completion rates across treatments is driven by the difference in the proportions of participants who never log any study hours after signing up for the experiment. This applies to nine of the 38 people (23.7%) in the weekly treatment, compared to two of the 36 people (5.6%) in the no-weekly treatment; this difference in rates is significant (Z = 2.19, p = 0.028, two-tailed test). This suggests that the realization that there were multiple hurdles may have deterred people from even starting the task.

<sup>&</sup>lt;sup>19</sup> This is reflected in the fact that both the weekly and no weekly requirements lines intersect with the 15-hour target line on Sundays – the last day of each week in the experiment.

without weekly requirements. Cumulatively, the study hours logged for winners were very close to a 'target' line of 15 hours per week.<sup>19</sup> The average number of study hours (75.35) for the winners was very close to the minimum of 75, ranging from 75.02 to 76.74. This suggests that students did not find this studying task to be innately pleasurable.



Figure 2

Is there a difference in study hours across weeks for those who completed the 75 hours of study? Table 1 presents the average number of hours for the winners by week and by group:

Week	Weekly requirements	No weekly requirements					
1	16.92 (0.41)	13.91 (0.50)					
2	14.53 (0.43)	16.08 (0.39)					
3	11.38 (0.50)	13.78 (0.44)					
4	15.75 (0.62)	16.27 (0.61)					
5	16.78 (0.69)	15.31 (0.71)					

 Table 1: Average weekly study hours (winners), by group

Note: Standard errors are in parentheses

There is no clear trend over time for either treatment. Regressions of hours against weeks yield insignificant coefficients for the time trends (0.09 and 0.30, respectively, with corresponding *t*-statistics of 0.11 and 0.76). This does not appear to be evidence of procrastination. However, if we look at the study patterns for each individual, there is evidence

that some people front-load studying hours while others delay.<sup>20</sup> Appendix A shows the study hours by individual, both for all participants and for only those people who completed the 75 hours. We classify people who finished as front-loaders (back-loaders) if they logged at least half of their hours in the first (last) two weeks of the five-week period. Of the 37 people who completed the task, four people front-loaded, while 11 people back-loaded.<sup>21,22</sup>

		Group	
Variables	(1)	(2)	(3)
	Weekly requirements	No weekly requirements	Pooled data
Week 2	-2.39 [1.57]	-2.39 2.17 [1.57] [1.41]	
Week 3	-5.54***	-0.13	-2.32*
	[1.62]	[1.36]	[1.11]
Week 4	-1.18	2.36	0.92
	[2.24]	[2.82]	[1.89]
Week 5	-0.14	1.40	0.78
	[2.82]	[2.93]	[2.04]
Constant	16.92***	13.91***	15.13***
	[1.15]	[1.36]	[0.95]
# Observations	75	110	185
$R^2$	0.12	0.02	0.03

Table 2 – Regressions for weekly studying hours (winners only)

Notes: Week 1 is the omitted variable in these regressions. Robust standard errors clustered by subject are in brackets. \* and \*\*\* indicate significance at the 10% and 1% level, respectively (two-tailed tests). The pooled regression includes a control for treatment group (not reported).

We present OLS regressions showing study patterns for both treatments in Table 2. The

regressions confirm that there is no clear increasing or decreasing profile over the course of the

<sup>&</sup>lt;sup>20</sup> In addition, one might classify those people who didn't finish (or didn't even start) as more serious procrastinators (we thank Jeroen van de Ven for this comment). Indeed, more extreme procrastinators meant to sign up, but didn't get around to it (or didn't even get around to submitting their college applications on time). <sup>21</sup> As some of the front-loaders may be people who study a great deal and who simply continued extensive studying

at home once 75 hours were logged, this is an upper bound on the number of front-loaders.

<sup>&</sup>lt;sup>22</sup> We see little difference across gender in this regard – one (three) of the 11 (26) male (female) winners chose to front-load, and three (eight) of the 11 (26) male winners chose to back-load; neither of these differences is close to statistical significance (Z = 0.22 and Z = 0.21 for the respective comparisons).

experiment. Only the coefficient for the dummy for Week 3 has any statistical significance, and this would appear to reflect the effect of a closure of the library during evening peak study time due to a power outage (see the dip around day 16 in Figure 2).

Table 3 shows the average hours of studying logged by winners on each day of the week. The number of study hours shows a tendency to decrease over the course of the week (from Monday to Thursday), with a dramatic drop on Friday and Saturday, and some recovery beginning Sunday afternoon. The patterns are essentially similar across the two treatments. Students appear to start the week fairly fresh and run out of steam as it progresses. The weekend appears to be the time when students 're-charge' (perhaps their willpower); alternatively, students might have a higher opportunity cost of studying on the weekend.

Day	Weekly requirements	No weekly requirements	Pooled data
Monday	3.14 (0.23)	3.25 (0.20)	3.21 (0.15)
Tuesday	2.69 (0.24)	2.94 (0.20)	2.84 (0.15)
Wednesday	3.04 (0.23)	2.87 (0.19)	2.94 (0.15)
Thursday	2.57 (0.30)	2.21 (0.22)	2.35 (0.18)
Friday	1.01 (0.17)	1.05 (0.15)	1.03 (0.11)
Saturday	1.06 (0.21)	0.90 (0.16)	0.97 (0.13)
Sunday	1.56 (0.24)	1.84 (0.24)	1.73 (0.17)

Table 3: Average study hours (winners), by day and group

Note: Standard errors are in parentheses

We see a strong and significant cyclical pattern, common to both treatments.<sup>23</sup> There is virtually no difference between the estimated coefficients for the two groups (when a dummy variable for group is added, its coefficient is 0.00; when we include week\*group interaction

<sup>&</sup>lt;sup>23</sup> Since the winners only represent 50 percent of the participants (37 of 74), it is natural to wonder whether there is more back-loading among those students who did not manage to complete the task. Naturally, the data for this group is far less complete, as most people who did not complete the studying task stopped logging hours early on. Appendix B shows a pattern somewhat similar to Figure 2 for the first two weeks of the study (70 percent of the people who did not complete the second week) for people who did not attain the studying target.

dummies, none of these has a coefficient that is close to statistical significance – the lowest *p*value is 0.47).

Although the regression results find no significant differences from Monday to Wednesday, analysis of each individual's logged study hours does suggest that people tend to log the most hours on Monday, with a steadily declining rate until it increases on Sunday. We can examine how many people logged the most and second-most study hours by day of the week (see Appendix C). For the full participant population, 22 of the 63 people (34.9%) who logged any study hours logged the highest number of study hours on Monday.<sup>24</sup> The pattern is even stronger if we only include the winners, with 15 of 37 people (40.5%) logging the highest number of study hours on Monday.<sup>25</sup> On the other hand, the non-winners have a higher proportion of study hours later in the week, with this being particularly true for the group with weekly restrictions; it seems plausible that these people tried to catch up late in the week, but did not succeed.

A final question of importance is whether completing (or even attempting) the study task was helpful in terms of performance. As mentioned earlier, 42 of our participants originated in an introductory class and we were given permission to access the (anonymous) grade records for the course, matching student ID numbers for the participants. There were quizzes, a midterm, and a final exam in the course. Our study commenced in the fourth week of the quarter, with two quizzes preceding our study. There is more variability in the quiz grades, with a number of people missing them, particularly after the midterm.<sup>26</sup> We therefore trust the midterm (taken after the first week of the five-week experiment) and final-exam (taken a week after the

<sup>&</sup>lt;sup>24</sup> This compares to 15 on Tuesday, 13 on Wednesday, six on Thursday, one on Friday, one on Saturday, and five on Sunday; the pattern is similar if we consider highest and second-highest days of the week. <sup>25</sup> This compares to six on Tuesday, eight on Wednesday, five on Thursday, and five on Sunday; again, the pattern is

similar if we consider highest and second-highest days of the week. <sup>26</sup> The absentee rate on the quizzes after the midterm was more than double the rate before the midterm.

experiment ended) scores more, but nevertheless include an average for the first two quizzes. Table 4 shows the mean scores by group:

Group	N	Quiz	Midterm	Final
Non-participants	403	3.13 (0.05)	8.78 (0.13)	7.56 (0.15)
Participants, non-winners	21	3.21 (0.16)	9.57 (0.51)	7.71 (0.4)
Participants, winners	21	3.67 (0.15)	10.71 (0.49)	9.24 (0.5)

Table 4: Mean scores on tests, by group

Note: Standard errors are in parentheses

We see that the differences between the non-participants and the non-winners are generally small, although slightly larger for the midterm.<sup>27,28</sup> In fact, Wilcoxon-Mann-Whitney (two-tailed) ranksum tests confirm that none of these differences are significant (for the midterm comparison, we find that Z = 1.37, p = 0.171). On the final exam, there was no difference between non-participants and non-winners (Z = 0.10). There is also no significant difference between non-winners and winners on the midterm scores (Z = 1.58, p = 0.115); however, the difference between final scores is in fact significant (Z = 2.38, p = 0.017). Thus, the difference in test scores across participants increased over the course of the quarter, suggesting that people who logged in 75 hours of library study improved their grades relative to those who signed up and did not complete the task.<sup>29</sup>

<sup>&</sup>lt;sup>27</sup> We note that the midterm took place before many non-winners had stopped logging study hours.

<sup>&</sup>lt;sup>28</sup> Given that only 42 of 445 students in these classes chose to participate, it may well be the case that there was a higher degree of procrastination amongst those students who did not participate. While we are primarily interested in treatment effects (and use random assignment to these treatments), we caution that the levels of the behavior observed may potentially not be representative of the overall student population.

<sup>&</sup>lt;sup>29</sup> One issue is whether the study hours in the monitored location were simply a substitute for study hours elsewhere. The data suggest that perhaps this is not completely the case. In addition, the results from our pre- and post-experiment questionnaires reveal that only 24% of the eventual winners studied more than 15 hours per week before the experiment started and 64% of winners reported reducing their weekly study hours once the experiment ended. This provides some evidence that the experiment increased total hours studied over the five-week period.

### Study 2

Of the 158 people who signed up online, 100 (63 percent) completed the task successfully. In a departure from our earlier results, 60 of 85 males (71 percent) completed this shorter task, compared to 40 of 73 females (55 percent); this difference is statistically significant (Z = 2.05, p = 0.040, two-tailed test). As might be expected, success rates were significantly higher for the Tuesday-Wednesday group, as 57 of 79 people (72 percent) in this group finished and 43 of the 79 people (54 percent) in the Friday-Saturday group finished (Z = 2.31, p = 0.021, two-tailed test).

Thirty-nine people answered all 20 questions on the first day, 15 people answered all 20 questions on the second day, and the other 46 people who completed the task answered questions on both days. Ninety-two people answered some questions on the first day, while 64 people answered questions on the second day. Forty-nine people never answered any questions, while the remaining nine people answered between four and 15 questions. Sixty-one of the 64 people (95 percent) who answered some questions on the second day completed the task successfully.

With respect to the people who completed the task, the average number of questions answered on the first (second) day was 11.53 (8.47); similarly, for the nine people who answered more than zero, but less than 20, questions in total, the average number of questions answered on the first (second) day was 5.56 (3.33).<sup>30</sup> We see some tendency for males to procrastinate more than females, as the number of questions answered on the first day for people who finished was 9.97 for males and 13.88 for females (Z = -2.25, p = 0.025, two-tailed test).

We observe differences in behavior according to the type of Stroop test that was assigned. Twenty-two of the 45 people (49 percent) who were assigned the concordant Stroop

<sup>&</sup>lt;sup>30</sup> Since more questions are answered on the first day, at first glance there is no evidence of procrastination in the aggregate. However, one must keep in mind that each participant was required to answer the Stroop tests on the first day, so that coming back on a second day involved an additional transaction cost.

and succeeded at the task answered all 20 questions on the first day, compared to 17 of the 55 people (31 percent) who completed the task and were assigned the discordant Stroop (Z = 1.83, p = 0.034, one-tailed test). This implies some degree of willpower depletion.<sup>31</sup> These results are combined in the regressions in Table 5 and the success rates by Stroop type and start day are displayed visually in Figure 3.

	independent variable						
Dependent Variables	(1) Success rate	(2) Questions, 1 <sup>st</sup> Day					
Tuesday	.539** (.212)	1.59 (3.34)					
Discordant Stroop	.406* (.212)	-6.92** (3.42)					
Male	.515** (.213)	-8.04** (3.45)					
Constant	384* (.222)	22.11*** (4.17)					
Observations	158	100					
Pseudo R <sup>2</sup>	0.068	0.018					
Notes: Standard errors in parentheses. Specification (1) is a Probit regression and (2) is a two-sided Tobit regression, including only those people who answered all 20 questions. *, **, and *** indicate significance at the 10%, 5%, and 1% level, respectively (two-tailed tests). We omit interaction terms, as these are not significant and do not qualitatively change the main results.							

 Table 5 – Determinants of success rate and number of questions answered

 Independent variable

<sup>&</sup>lt;sup>31</sup> We are not wedded to the term "willpower", as this could be seen as, for example, depletion of cognitive resources. We use this term simply to reflect the standard terminology in the psychology literature.



Figure 3 - Success rates, by Stroop type and start day

A rather surprising result is that people who were assigned the discordant Stroop were actually slightly *more* likely to eventually finish the task. Fifty-five of the 79 people (70 percent) assigned the discordant Stroop finished, while 45 of the 79 people (57 percent) assigned the concordant Stroop finished; the test of proportions gives Z = 1.65, p = 0.099, two-tailed test. Specification (2) confirms that people who completed the task after performing the discordant Stroop answered significantly fewer questions on the first day, and that males who completed the task delayed more than did females.

Another measure that may reflect willpower depletion, but in any case certainly reflects the quality of work, is the percentage of questions answered correctly. There is no significant difference in the percentage correct across gender (62 percent for males vs. 60 percent for females) or type of Stroop test (60 percent in both cases). In addition, the percentage of correct answers on the first day does not predict success rates (the *t*-statistic on the coefficient for number of questions is 0.21). However we do find that the percentage of correct answers on the

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first day is significantly higher than on the second day.<sup>32</sup> This is robust to whether we consider the whole sample of 158 people, the 100 people who finished, or the 47 people who answered questions on both days.<sup>33</sup> The percentage of correct answers is also higher for the Tuesday-Wednesday group (0.63 vs. 0.56, Z = 1.78, p = 0.038, one-tailed test).

One of the features of our data is the pronounced weekly studying cycles in Study 1 and the day-of-the-week effects in Study 2. Although Study 2 (and a number of studies in the psychology literature) provides evidence that willpower is a resource that can be depleted, it is not clear whether the weekly cycles in Study 1 reflect the depletion of willpower, changes in the opportunity cost of studying or some other influence. Nevertheless, some data from Study 1 offer some insight into this question. The opportunity-cost story would be most consistent with a difference in behavior during the workweek and during the weekend, rather than a daily difference (at least during the workweek). On this basis, we should expect similar behavior for the period from Monday through Thursday. The individual data suggest that Monday is the favorite day for logging study hours, although we do not find significant differences for Monday-Wednesday. In addition, we see a significant decline in study hours on Thursday.<sup>34</sup>

### Questionnaire

<sup>&</sup>lt;sup>32</sup> The order of questions was randomized, so there should be no difference in difficulty levels across days. One possible explanation for the lower success rate on the second day is that students may have "sprinted to the finish" once the finish was within sight.

<sup>&</sup>lt;sup>33</sup> The respective comparisons are 0.66 vs. 0.49 (Z = 4.35, p = 0.000), 0.66 vs. 0.50 (Z = 4.10, p = 0.001), and 0.67 vs. 0.47 (Z = 3.43, p = 0.001). All of these tests are two-tailed.

<sup>&</sup>lt;sup>34</sup> While some might argue that the weekend starts on Thursday evening, time-of-day data (shown in Appendix E) show more of a decrease (relative to Wednesday) for Thursday morning and afternoon, rather than Thursday evening; the percentage decrease in logged study hours for the morning and afternoon combined is 36.4%, while this percentage decrease is only 14.2% for the evening. While this suggests that willpower is being depleted through the workweek, it is hardly conclusive.

<sup>&</sup>lt;sup>36</sup> While both sets of numbers reflect the prevailing wage and are much higher than students in Study 1 were paid per hour, we are primarily interested in the relative costs. We report medians rather than means to avoid distortions from outliers.

The main question in the survey concerns how people would allocate 75 hours of library study time over each week of a 5-week period. In one version, people were asked how they would <u>prefer</u> to allocate their time during this period, while another version asked how they thought they would <u>end up</u> allocating their time during this period. The results are shown in Table 6:

Week	Optimal	Expected
1	13.85 (1.37)	13.56 (1.20)
2	15.45 (1.07)	12.88 (1.00)
3	16.30 (1.08)	14.15 (0.91)
4	13.67 (1.07)	15.85 (1.04)
5	15.24 (2.17)	18.85 (1.92)

Table 6: Optimal and expected weekly study hours, questionnaire

Note: Standard errors are in parentheses.

The aggregate optimal studying profile is essentially flat, with the hours for each week never significantly different from 15. The aggregate expected studying profile does show evidence that people expect to delay somewhat in the task. A second question asked how people would allocate 15 hours of library study over the days of the week. Here the pattern resembles that seen in Table 3, with 2.82, 2.59, 2.87, 1.81, 0.91, 1.08, and 2.90 hours projected respectively for Monday – Sunday. Thus, people do appear to be aware of their actual intra-week preferences, as only the expected hours for Sunday study differ much from the actual numbers in Table 3.

We also asked students about how likely it was that they would not be able to complete a project requiring periodic work over a 5-week period; this is a proxy for the degree of uncertainty in the students' lives. Overall, there is only a modest degree of uncertainty, with a median response of 10%. Two additional questions examined the convexity of effort costs. In one question, students were asked for the minimum amount of money they would require to study in the library for one, two, or four hours in a day; the median responses were \$8, \$18, and \$40, respectively, indicating a slight degree of convexity. A second question asked students for

the minimum amount of money they would require to study in the library for 10, 20, or 30 hours during a one-week period; the median responses were \$55, \$120, and \$300. Both sets of responses indicate at least a moderate degree of convexity in effort cost.<sup>36</sup>

Finally, we asked a question bearing on Study 2, in which the total amount of time required for the task was about 90 minutes. Students were asked whether they would rather spend 90 minutes on a task in one sitting at home or to spread the task out over two sittings (one on each of two consecutive days) at home, with a total time of either 90, 105, or 120 minutes. Forty-six people always preferred to have only one sitting, while 10 people always preferred two sittings.<sup>37</sup> Thus, for a high percentage of the students (67%) the convexity of effort cost was not sufficient to overcome the additional transactions cost of the additional sitting.

### 4. **DISCUSSION**

### 4.1. Related behavioral models

A recent theoretical literature in economics has explored issues of procrastination, willpower and time-inconsistency. We briefly summarize several models and explore their broad predictions in our design and how consistent these are with our data. Although our research objective was to test if (i) interim deadlines help task completion, (ii) willpower depletion hurts task completion, and not to prove/disprove various theoretical explanations for time-inconsistency, our results do shed empirical light on the different modeling approaches.

One notion is that people have present-biased preferences. A person puts more weight on

<sup>&</sup>lt;sup>37</sup> Only three of the other sets of responses made sense to us, with students preferring two sittings except when this would take 120 minutes. The other 10 people appear to have been confused, as six of these preferred one sitting except when two sittings would take 105 or 120 minutes, and four of these preferred one sitting except when two sittings would take 120 minutes.

the present than on the future, leading to dynamic inconsistency.<sup>38</sup> Strotz (1955), Ainslie (1992), Laibson (1997) and O'Donoghue and Rabin (1999) discuss present-biased (quasi-hyperbolic) preferences as an explanation for persistent bad habits and addictions.<sup>39</sup> In Study 1, the model predicts that with weekly requirements, study hours should be logged near the end of each week; without weekly requirements, there should be a surge in hours logged late in the five-week period. However, since uncertainty could potentially overwhelm the tendency to procrastinate, the predictions for the studying profile in Study 1 will depend on assumptions about parameter values; we perform a calibration exercise using the data from our questionnaire. There are no particular predictions in Study 2; certainly the model does not predict the higher overall success rate after exposure to the discordant Stroop test.

Another notion relates to temptation and commitment. The Gul and Pesendorfer (2001) model suggests that commitments may be helpful for overcoming the temptation to avoid an unattractive task in the future. While the predictions depend on parameter values, if we view weekly restrictions as reducing the choice set of the agent, so that there is effectively less temptation, one might expect a positive impact on performance. The predictions in Study 2 are less clear; however, if an agent who completes the discordant Stroop views this accomplishment as a commitment to complete the task eventually, the model would predict less temptation on the second day and so a higher completion rate overall.

<sup>&</sup>lt;sup>38</sup> Overall, one of the causes for apparent reversals in preferences over time seems to be the change in the saliency of the costs and benefits of the activity in question (Akerlof 1991).

<sup>&</sup>lt;sup>39</sup> Frederick, Loewenstein, and O'Donoghue (2002) provide a comprehensive review of empirical research on intertemporal choice, as well as an overview of related theoretical models. We would also like to mention two very new papers. Bisin and Hyndman (2008) investigate stopping-time problems and characterize behavior for exponential, naïve-hyperbolic and sophisticated-hyperbolic discounters. They show that an agent with standard time preferences who suffers from "temptation and self-control" would never be willing to self-impose a deadline. Suvorov and van de Ven (2008) develop a theory of self-regulation based on goal setting. They derive a condition under which proximal short-term goals are better than distal long-term goals.

Fudenberg and Levine (2006) propose a dual-self model in which a long-run self effectively ties the hands of a series of myopic short-run selves, serving as a form of commitment.<sup>40</sup> In the first phase, the sophisticated long-run self chooses a commitment action that affects any short-run self's utility; in the second phase, the short-run self then chooses the action (e.g., how much to spend). In a similar spirit to the Gul and Pesendorfer (2001) model, if the restrictions in Study 1 serve to reduce the self-control problem by reducing the set of options, restrictions could be beneficial. Since "increased cognitive load makes temptations harder to resist" in this model, it predicts that students who are assigned the discordant Stroop in Study 2 should procrastinate and presumably be less successful.

Two other models instead consider self-control problems in the context of strength of will or willpower.<sup>41</sup> Bénabou and Tirole (2004) consider a model in which self-signaling leads to self-regulation; the key is that people have imperfect knowledge of their *strength of will*, learning it only through experience and having difficulty later remembering it. If one decides to attempt a willpower activity in the first sub-period, he or she is faced with a (stronger) temptation in the second sub-period and experiences a craving cost. Since people have only imperfect knowledge of their own willpower, their choices serve as 'signals' about their types; a negative signal (precedent) undermines self-confidence and willpower. While the predictions depend on parameter values, one aspect of the model is that "the degree of self-control an individual can achieve is shown to … decrease with prior external constraints." A natural interpretation of this result in our context is that a participant with weekly requirements in Study 1 will tend to be less likely to complete the studying task. In Study 2, if the discordant Stroop test enhances a

<sup>&</sup>lt;sup>40</sup> Thaler and Shefrin (1981) present an early agency model of self-control, involving a farsighted *planner* and a myopic *doer*.

<sup>&</sup>lt;sup>41</sup> These models relate closely to the experimental work by Baumeister, Heatherton, and Tice (1994) and Baumeister and Vohs (2003), which finds that people who have just been required to exercise self-restraint tend to exhibit a lower degree of self-control.

student's recall of their own strength of will, then students subjected to this test will tend to be more successful on the second day.<sup>42</sup>

Ozdenoren, Salant, and Silverman (forthcoming) provide an explanatory model based on willpower as a depletable (but renewable) resource. A willpower-constrained individual will regard seemingly unrelated activities as linked because he or she uses the same cognitive resources to exercise self-control in different activities. If willpower is depleted during one part of the week and replenished during another, we could see weekly studying cycles in Study 1. Weekly requirements will tend to induce better performance, since requirments substitute for using willpower. In Study 2, students subjected to the discordant Stroop should tend to postpone answering questions until the second day. We would expect more success for the weekday group, as their willpower is presumably higher than the weekend group. However, this model does not predict a higher success rate for those people who experienced the discordant Stroop.

To summarize our discussion, the predictions of the behavioral models often depend on parameter values, many of which cannot be readily determined. In any case, none of the models as currently constituted can explain all of our data. While we observe some people back-loading in Study 1, the aggregate weekly profile over time of logged library study hours is fairly flat, providing little evidence of procrastination. While this profile might imply that these students do not have present-biased preferences, the profile is entirely consistent with a mild degree of present bias if we incorporate the modest degree of uncertainty we find in our questionnaire.

Although Study 2 (and a number of studies in the psychology literature) provides evidence that willpower is a resource that can be depleted, it is not clear whether the weekly

<sup>&</sup>lt;sup>42</sup> Of course, alternative explanations also exist; for example, one could see having experienced the discordant Stroop as a psychological 'sunk cost'.

cycles in Study 1 reflect the depletion of willpower, changes in the opportunity cost of studying or some other influence. Nevertheless, the opportunity-cost story would be most consistent with a difference in behavior during the workweek and during the weekend, rather than a daily difference during the workweek. The individual data suggest that Monday is the favorite day for logging study hours, even though we do not find significant differences for Monday-Wednesday in our regression analysis. In addition, we see a significant decline in study hours on Thursday.

Regarding our behavioral interventions, we expected that students would procrastinate and that weekly requirements would mitigate this tendency. If the weekly restrictions in Study 1 do not help much with self-control problems, the additional hurdles imposed should make success less likely in this case. Study 2 provides evidence of the importance of willpower in self-control problems. Finally, the effect in Study 2 of the discordant Stroop on performance on the first day is consistent with all models that permit self-control to be linked with willpower or cognitive load, with these data most in keeping with willpower models.

### 4.2 A calibration exercise

We use the results on uncertainty and the cost of studying from the questionnaire to simulate the expected behavior for agents with different values of  $\beta$  (either 1.0, 0.9, or 0.7) with and without uncertainty over the cost of future studying.<sup>43</sup> Appendix F shows how we calibrate the cost function for study hours in a week; we describe how we estimate the weekly profile for various parameter combinations and present the simulations in Appendix G.

$$\forall t, U^t(u_t, u_{t+1}, \dots, u_T) \equiv \delta^t u_t + \beta \sum_{\tau=t+1}^t \delta^\tau u_{\tau}$$
, where  $0 < \beta$  and  $\delta \le 1$ 

<sup>&</sup>lt;sup>43</sup> The quasi-hyperbolic formulation from Laibson (1997) and O'Donoghue and Rabin (1999) for preferences across a stream of utilities, where  $u_t$  is a person's instantaneous utility in period *t*, is:

Here  $\delta$  is the standard, time-consistent, exponential discount rate and  $\beta < 1$  indicates a "bias for the present".

We see that the calibrated cost function predicts a weekly profile that is essentially flat for an exponential discounter ( $\beta = 1$ ) without uncertainty. This pattern (which we observe in the aggregate) is also predicted for an agent with mild time-inconsistency ( $\beta = 0.9$ , as suggested by Shapiro, 2005 and Meier and Sprenger, 2010) and uncertainty. However, when we instead use the value of  $\beta = 0.7$ , as suggested by McClure *et al.* (2007) and Laibson *et al.* (2007), the model predicts that even with uncertainty, studying hours increase sharply over time. Thus, while the aggregate pattern that we observe is not fundamentally inconsistent with the quasi-hyperbolic models, it appears to rule out population values for  $\beta$  that are substantially less than 1.

### 4.3. Penalties and behavior

We now wish to discuss the differences between the results in Study1 and those found in Ariely and Wertenbroch (2002). Some critical differences include the likelihood of completion without any deadlines, the threshold nature of the interim deadlines, and the 'soft' penalties in their study. For those people who are in a class at MIT, not turning in a term paper will result in an extremely bad outcome; thus, everyone completes the overall task, with performance (but not completion) affected by the characteristics of the various deadlines. In their proofreading study, one can turn in the project at any point, regardless of whether one has finished the task; payment only depends on the amount and quality of the work done, with a small penalty for being late. By comparison, in Study 1 the consequence of failure is simply not receiving \$95, but there are no shades of gray involved. We had expected that this strict penalty would serve as a stronger incentive for people to avoid procrastination. While Ariely and Wertenbroch (2002) provide an 'existence proof' that deadlines can be beneficial,<sup>44</sup> our findings suggest that this is not always

<sup>&</sup>lt;sup>44</sup> However, Amir and Ariely (2008) find that discrete progress markers may "generate complacency, sway motivation away from the end goal, and decrease performance in the task" when the distance to the goal is certain, as is the case in Study 1.

the case.

Of course, these are entirely different environments. An open question is which pertains more to the field. In many environments where procrastination may matter a lot, there is a more continuous relationship between effort and performance. For example, a student's grade on an exam will typically not be either 0 or 100 depending on the degree of procrastination; however, it may well determine whether the student passes or fails a course. Similarly, contracts may contain performance penalties for being late. On the other hand, whether or not one misses one's flight is a binary outcome, as is (in most cases) a submission deadline for a conference or a special issue. This is even true in sports, as a professional golfer who misses a tee time will be disqualified. Our view is that both the Ariely and Wertenbroch (2002) framework and our own are representative of important economic settings, so that it is important to delineate effects in each case, as well as in others.<sup>45</sup> While the Ariely and Wertenbroch study is fine to establish the existence of benefits from interim deadlines, but it may not help much when it comes to identifying parameters; while we could re-run our study with a \$80 (or \$150) payoff, but it seems difficult to exogenously adjust their approach.

To summarize our discussion, the predictions of the behavioral models often depend on parameter values, many of which cannot be readily determined. In any case, none of the models as currently constituted can explain all of our data. First, there is no explicit explanation in these models for the strong day-of-the-week effect found in both Study 1 and Study 2, although to the extent that this reflects differences in opportunity costs, enhanced versions of the models could incorporate this feature. Second, while we observe some people back-loading in Study 1, the

<sup>&</sup>lt;sup>45</sup> While the Ariely and Wertenbroch study is fine to establish the existence of benefits from interim deadlines, but it may not help much when it comes to identifying parameters; while we could re-run our study with a \$80 (or \$150) payoff, but it seems difficult to exogenously adjust their approach.

aggregate weekly profile over time of logged library study hours is fairly flat, providing little evidence of procrastination. While this profile might imply that these students do not have present-biased preferences, the profile is entirely consistent with a mild degree of present bias if we incorporate the modest degree of uncertainty we find in our questionnaire. Third, Study 2 provides evidence of the importance of willpower in self-control problems.

The behavioral interventions that we attempted led to results that surprised us to some degree. In Study 1, we thought that well-chosen deadlines would help people to overcome their tendency to procrastinate and be more likely to complete the studying task than people not facing restrictions; however, we find the opposite result. In Study 2, while we do find that the discordant Stroop tends to induce people to delay more of the task until the second day, the overall success rate is actually higher for the people who faced the willpower-depleting Stroop on the first day.

### 5. CONCLUSION

Our study is one of the first to provide empirical evidence concerning procrastination, willpower, and monetary incentives in tasks involving duration. Since it is exploratory work, we did not expect to fully resolve all of the issues present in this regard. Nevertheless, we identify a number of interesting patterns, provide data and feedback in relation to the relevant theoretical models, and suggest follow-up studies to other researchers. We find evidence of willpower depletion, as well as behavior suggestive of self-reputation considerations. The limited evidence of back-loading suggests that the overall degree of present bias in our sample is not severe. Our sense is that procrastination is not a simple phenomenon and none of the current models fully capture the mechanisms through which procrastination occurs.

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A number of puzzles remain. For example, while the discordant Stroop exercise did deplete willpower on the first day, a natural prediction would seem to be that this depletion would reduce overall success rates. Yet, overall success rates were higher for the willpowerdepleted group. Is exerting willpower self-signaling or does this result represent some sort of commitment due to having suffered through this exercise, inducing the determination to complete the task? How much domain specificity or crossover from cognitive load is there, in the sense of Ozdenoren, Salant, and Silverman (forthcoming) or Fudenberg and Levine (2006)? Have we identified differences in procrastination and willpower across gender, particularly in terms of differences according to whether a task is long or short in duration?

The fact that the behavioral interventions in both studies were ineffective points out the lack of existing data on how incentives affect behavior on tasks involving significant duration. Our Study 1 results suggest that commitment devices (at least externally-imposed ones) may in fact be counter-productive if they involve imposing additional hurdles to overcome. That our studying task and the tasks considered by Ariely and Wertenbroch (2002) produce decidedly different results suggests that the tendency to procrastinate varies significantly with incentives and context. Further, in relation to the effectiveness of commitment devices, we strongly suspect that details matter. In some cases, there may be a preference for flexibility while in other cases there may be a preference for commitment. The issue of how to overcome self-control problems has important economic consequences and it seems that there is not a simple answer.

It is quite clear that our results are only a beginning. We consider the area of time structuring, procrastination, and incentives to be just coming into its first full flowering. We hope that researchers in this area can build upon our results in conducting both more empirical research and more theoretical work. Our studies could provide motivation for the design of other

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studies aimed at mapping out the boundaries of effective deadlines, and could form the starting

point for designing a series of studies to test the various behavioral theories mentioned, as well

as others yet to be formulated.

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## **APPENDIX A – INDIVIDUAL STUDY HOURS IN STUDY 1**

ID	Female	Week 1	Week 2	Week 3	Week 4	Week 5	Total
1	1	0	0	0	0	0	0.00
2	1	12.31	20.15	13.82	14.44	14.40	75.12
3	0	9.18	14.78	4.12	1.53	0	29.61
4	0	11.16	15.16	16.89	17.92	14.30	75.42
5	1	8.17	9.97	2.46	9.26	0	29.85
6	0	16.70	13.52	6.66	17.47	20.90	75.25
7	1	1.83	0	0	0	0	1.83
8	1	2.15	18.75	4.35	31.19	18.63	75.07
9	1	9.34	6.77	0	2.74	14.31	33.17
10	1	10.27	0.67	3.43	0	0	14.36
11	1	17.72	17.74	20.98	10.01	8.74	75.18
12	1	8.69	18.11	17.43	10.24	20.64	75.11
13	1	12.13	12.43	5.62	5.03	39.93	75.13
14	1	2.82	0	0	0	0	2.82
15	0	18.11	11.99	12.42	18.20	14.49	75.21
16	1	12.54	13.74	10.16	19.61	19.26	75.30
17	0	14.14	22.03	18.05	14.30	7.27	75.80
18	1	0	12.87	13.76	2.74	0	29.38
19	0	16.37	6.58	27.47	8.33	16.63	75.37
44	1	13.69	9.77	13.50	13.05	25.07	75.08
45	1	14.98	12.14	16.97	22.90	8.21	75.19
46	1	24.56	28.81	16.97	5.46	0	75.80
47	1	17.68	13.94	11.94	15.77	16.56	75.89
48	1	1.33	0	0	0	0	1.33
49	1	3.81	14.67	12.08	27.67	17.48	75.72
50	1	16.32	15.49	9.53	14.44	19.38	75.16
51	1	15.19	17.00	11.51	14.79	16.79	75.27
52	1	6.20	11.79	13.74	23.09	20.56	75.38
53	1	5.71	15.80	6.40	29.69	17.58	75.17
54	1	28.93	24.16	19.03	3.33	0	75.46
55	0	1.07	3.79	0	0	0	4.87
56	1	0	0	0	0	0	0.00
57	0	12.93	0.99	0	0	0	13.92
58	0	10.64	0	3.53	0	0	14.17
59	0	7.60	0	0	0	0	7.60
60	1	16.93	19.94	17.73	20.99	0	75.60

## WEEKLY STUDY HOURS IN NO-WEEKLY GROUP, BY INDIVIDUAL

ID	Female	Week 1	Week 2	Week 3	Week 4	Week 5	Total
20	0	0	0	0	0	0	0.00
21	0	0	0	0	0	0	0.00
22	0	12.34	11.62	0.02	0	0	23.98
23	0	19.07	15.41	15.86	13.10	13.29	76.74
24	0	14.97	17.59	8.66	12.72	21.11	75.05
25	0	0	1.52	0	0	0	1.52
26	1	19.70	14.12	12.74	7.08	21.47	75.11
27	1	13.84	11.67	11.12	7.40	0	44.03
28	0	14.70	13.85	8.35	27.88	11.19	75.97
29	0	12.22	3.21	0	0	0	15.42
30	0	0	0	0	0	0	0.00
31	1	18.02	8.23	6.65	15.43	23.89	72.21
32	1	0	0	0	0	0	0.00
33	0	30.06	16.99	12.86	15.11	0	75.02
34	0	12.02	12.07	12.21	22.59	16.24	75.13
35	1	0	0	0	0	0	0.00
36	1	13.53	13.85	19.62	10.30	17.79	75.09
37	0	0	0	0	0	0	0.00
38	1	0	0	0	0	0	0.00
39	0	0	0	0	0	0	1.73
40	0	0	0	0	0	0	0.00
41	0	11.81	4.52	0	0	0	16.33
42	1	5.31	0	0	0	0	5.31
43	1	15.55	13.10	10.23	16.69	20.50	76.06
61	0	12.43	3.82	0	0	0	16.25
62	1	13.02	15.41	10.49	12.42	23.80	75.14
63	1	15.44	28.28	4.82	9.00	17.59	75.13
64	1	14.81	13.60	9.98	13.20	23.67	75.26
65	1	0	0	0	0	0	0.00
66	0	1.41	0	0	0	0	1.41
67	1	0.95	0	0	0	0	0.95
68	1	12.53	12.32	11.93	16.05	2.34	55.18
69	0	14.40	1.61	0	0	0	16.01
70	1	15.60	12.30	7.70	16.07	6.28	57.95
71	0	17.40	7.24	20.59	24.37	5.44	75.04
72	1	19.58	15.91	6.25	21.57	12.07	75.37
73	1	18.36	11.05	16.21	7.83	21.57	75.02
74	1	13.22	11.87	1.77	22.24	26.02	75.11

# WEEKLY STUDY HOURS IN WEEKLY GROUP, BY INDIVIDUAL

ID	Female	Week 1	Week 2	Week 3	Week 4	Week 5	Total
2	1	12.31	20.15	13.82	14.44	14.40	75.12
4	0	11.16	15.16	16.89	17.92	14.30	75.42
6	0	16.70	13.52	6.66	17.47	20.90	75.25
8	1	2.15	18.75	4.35	31.19	18.63	75.07
11	1	17.72	17.74	20.98	10.01	8.74	75.18
12	1	8.69	18.11	17.43	10.24	20.64	75.11
13	1	12.13	12.43	5.62	5.03	39.93	75.13
15	0	18.11	11.99	12.42	18.20	14.49	75.21
16	1	12.54	13.74	10.16	19.61	19.26	75.30
17	0	14.14	22.03	18.05	14.30	7.27	75.80
19	0	16.37	6.58	27.47	8.33	16.63	75.37
44	1	13.69	9.77	13.50	13.05	25.07	75.08
45	1	14.98	12.14	16.97	22.90	8.21	75.19
46	1	24.56	28.81	16.97	5.46	0	75.80
47	1	17.68	13.94	11.94	15.77	16.56	75.89
49	1	3.81	14.67	12.08	27.67	17.48	75.72
50	1	16.32	15.49	9.53	14.44	19.38	75.16
51	1	15.19	17.00	11.51	14.79	16.79	75.27
52	1	6.20	11.79	13.74	23.09	20.56	75.38
53	1	5.71	15.80	6.40	29.69	17.58	75.17
54	1	28.93	24.16	19.03	3.33	0	75.46
60	1	16.93	19.94	17.73	20.99	0	75.60

## WEEKLY STUDY HOURS IN NO-WEEKLY GROUP, BY INDIVIDUAL (WINNERS)

## WEEKLY STUDY HOURS IN WEEKLY GROUP, BY INDIVIDUAL (WINNERS)

ID	Female	Week 1	Week 2	Week 3	Week 4	Week 5	Total
23	0	19.07	15.41	15.86	13.10	13.29	76.74
24	0	14.97	17.59	8.66	12.72	21.11	75.05
26	1	19.70	14.12	12.74	7.08	21.47	75.11
28	0	14.70	13.85	8.35	27.88	11.19	75.97
33	0	30.06	16.99	12.86	15.11	0	75.02
34	0	12.02	12.07	12.21	22.59	16.24	75.13
36	1	13.53	13.85	19.62	10.30	17.79	75.09
43	1	15.55	13.10	10.23	16.69	20.50	76.06
62	1	13.02	15.41	10.49	12.42	23.80	75.14
63	1	15.44	28.28	4.82	9.00	17.59	75.13
64	1	14.81	13.60	9.98	13.20	23.67	75.26
71	0	17.40	7.24	20.59	24.37	5.44	75.04
72	1	19.58	15.91	6.25	21.57	12.07	75.37
73	1	18.36	11.05	16.21	7.83	21.57	75.02
74	1	13.22	11.87	1.77	22.24	26.02	75.11



**APPENDIX B** - Patterns In Study Hours, Study 1 Non-Winners

## **APPENDIX C – Individual Day of Week Data**

Day of week	# of times most	# of times second	# of times $1^{st}$ or $2^{nd}$
Monday	11 (32.4%)	9 (29.0%)	20 (58.8%)
Tuesday	9 (26.5%)	9 (29.0%)	18 (52.9%)
Wednesday	9 (26.5%)	5 (16.1%)	14 (41.2%)
Thursday	3 (8.8%)	4 (12.9%)	7 (20.6%)
Friday	0 (0.0%)	0 (0.0%)	0 (0.0%)
Saturday	0 (0.0%)	1 (3.2%)	1 (2.9%)
Sunday	2 (5.9%)	3 (9.7%)	5 (14.7%)

## No-weekly group, all

36 people in this group; 2 people logged no hours, 3 others only logged hours on 1 day

# Weekly group, all

Day of week	# of times most	# of times second	# of times 1 <sup>st</sup> or 2 <sup>nd</sup>
Monday	11 (37.9%)	4 (16.0%)	15 (51.7%)
Tuesday	6 (20.7%)	4 (16.0%)	10 (34.5%)
Wednesday	4 (13.8%)	7 (28.0%)	11 (37.9%)
Thursday	3 (10.3%)	1 (4.0%)	4 (13.8%)
Friday	1 (3.4%)	2 (8.0%)	3 (10.3%)
Saturday	1 (3.4%)	3 (12.0%)	4 (13.8%)
Sunday	3 (10.3%)	4 (16.0%)	7 (24.1%)

38 people in this group; 9 people logged no hours, 4 others only logged hours on 1 day

## Combined, all

Day of week	# of times most	# of times second	# of times 1 <sup>st</sup> or 2 <sup>nd</sup>
Monday	22 (34.9%)	14 (25.0%)	36 (57.1%)
Tuesday	15 (23.8%)	13 (23.2%)	28 (44.4%)
Wednesday	13 (20.6%)	12 (21.4%)	25 (39.7%)
Thursday	6 (9.5%)	4 (7.1%)	10 (15.9%)
Friday	1 (1.6%)	2 (3.6%)	3 (4.8%)
Saturday	1 (1.6%)	4 (7.1%)	5 (7.9%)
Sunday	5 (7.9%)	7 (12.5%)	7 (11.1%)

74 people in total; 11 people logged no hours, 7 others only logged hours on 1 day

# No-weekly group, winners

Day of week	# of times most	# of times second	# of times 1 <sup>st</sup> or 2 <sup>nd</sup>
Monday	7 (31.8%)	8 (36.4%)	15 (68.2%)
Tuesday	4 (18.2%)	8 (36.4%)	12 (54.5%)
Wednesday	6 (27.3%)	3 (13.6%)	9 (40.9%)
Thursday	3 (13.6%)	1 (4.5%)	4 (18.2%)
Friday	0 (0.0%)	0 (0.0%)	0 (0.0%)
Saturday	0 (0.0%)	0 (0.0%)	0 (0.0%)
Sunday	2 (9.1%)	2 (9.1%)	4 (18.2%)

22 people in this group

# Weekly group, winners

Day of week	# of times most	# of times second	# of times $1^{st}$ or $2^{nd}$
Monday	8 (53.3%)	2 (13.3%)	10 (66.7%)
Tuesday	2 (13.3%)	4 (26.7%)	6 (40.0%)
Wednesday	2 (13.3%)	5 (33.3%)	7 (46.7%)
Thursday	2 (13.3%)	1 (6.7%)	3 (20.0%)
Friday	0 (0.0%)	0 (0.0%)	0 (0.0%)
Saturday	0 (0.0%)	2 (13.3%)	2 (13.3%)
Sunday	1 (6.7%)	1 (6.7%)	2 (13.3%)

15 people in this group

# Combined, winners

Day of week	# of times most	# of times second	# of times $1^{st}$ or $2^{nd}$
Monday	15 (40.5%)	10 (27.0%)	25 (67.6%)
Tuesday	6 (16.2%)	12 (32.4%)	18 (48.6%)
Wednesday	8 (21.6%)	8 (21.6%)	16 (43.2%)
Thursday	5 (13.5%)	2 (5.4%)	7 (18.9%)
Friday	0 (0.0%)	0 (0.0%)	0 (0.0%)
Saturday	0 (0.0%)	2 (5.4%)	2 (5.4%)
Sunday	3 (8.1%)	3 (8.1%)	6 (16.2%)

37 people in total

# No-weekly group, non-winners

Day of week	# of times most	# of times second	# of times 1 <sup>st</sup> or 2 <sup>nd</sup>
Monday	4 (33.3%)	1 (11.1%)	5 (41.7%)
Tuesday	5 (41.7%)	1 (11.1%)	6 (50.0%)
Wednesday	3 (25.0%)	2 (22.2%)	5 (41.7%)
Thursday	0 (0.0%)	3 (33.3%)	3 (25.0%)
Friday	0 (0.0%)	0 (0.0%)	0 (0.0%)
Saturday	0 (0.0%)	1 (11.1%)	1 (8.3%)
Sunday	0 (0.0%)	1 (11.1%)	1 (8.3%)

14 people in this group; 2 people logged no hours, 3 others only logged hours on 1 day

# Weekly group, non-winners

Day of week	# of times most	# of times second	# of times 1 <sup>st</sup> or 2 <sup>nd</sup>
Monday	3 (21.4%)	2 (20.0%)	5 (35.7%)
Tuesday	4 (28.6%)	0 (0.0%)	4 (28.6%)
Wednesday	2 (14.3%)	2 (20.0%)	4 (28.6%)
Thursday	1 (7.1%)	0 (0.0%)	1 (7.1%)
Friday	1 (7.1%)	2 (20.0%)	3 (21.4%)
Saturday	1 (7.1%)	1 (10.0%)	2 (14.3%)
Sunday	2 (14.3%)	3 (30.0%)	5 (35.7%)

23 people in this group; 9 people logged no hours, 4 others only logged hours on 1 day

## Combined, non-winners

Day of week	# of times most	# of times second	# of times 1 <sup>st</sup> or 2 <sup>nd</sup>
Monday	7 (26.9%)	3 (15.8%)	10 (38.5%)
Tuesday	9 (34.6%)	1 (5.3%)	10 (38.5%)
Wednesday	5 (19.2%)	4 (21.1%)	9 (34.6%)
Thursday	1 (3.8%)	3 (15.8%)	4 (15.4%)
Friday	1 (3.8%)	2 (10.5%)	3 (11.5%)
Saturday	1 (3.8%)	2 (10.5%)	3 (11.5%)
Sunday	2 (7.7%)	4 (21.1%)	6 (23.1%)

37 people in total; 11 people logged no hours, 7 others only logged hours on 1 day

### **APPENDIX D: Questionnaire**

Suppose that you are getting paid to study a total of 75 hours (in a room in the library) over a five-week period, but you will receive no payment if you don't complete the 75 hours. Assuming that you plan to complete this study, *how would you prefer to allocate your time during this period* [in version 2: *how do you think you would end up allocating your time during this period*] in terms of hours per week? Take into account your studying preferences and your other activities during an average quarter.

Week	Hours you would study
1	
2	
3	
4	
5	

2) If you committed to complete a class project that required periodic work each week for five weeks, how likely do you think it is that you <u>wouldn't</u> complete it due to unexpected events in your life?

 0%
 10%
 20%
 30%
 40%
 50%
 60%
 70%
 80%
 90%
 100%
 Definitely complete

3) For this question, assume that you are doing homework at home.

Would you rather do **1 hour and 30 minutes** of homework at one sitting at home (with short breaks permitted), or instead spread out **1 hour and 30 minutes** of homework over two sittings, one on each of two days?

Do 1 hour and 30 minutes in one sitting Spread 1 hour and 30 minutes over two sittings

Would you rather do **1 hour and 30 minutes** of homework at one sitting at home (with short breaks permitted), or spread out **1 hour and 45 minutes** of homework over two sittings, one on each of two days?

Do 1 hour and 30 minutes in one sitting Spread 1 hour and 45 minutes over two sittings

Would you rather do **1 hour and 30 minutes** of homework at one sitting at home (with short breaks permitted), or spread out **2 hours** of homework over two sittings, one on each of two days?

Do 1 hour and 30 minutes in one sitting

Spread 2 hours over two sittings

4) Suppose that you planned to study 15 hours in a particular week. How would you allocate this time over the week, in terms of hours on each of Monday, Tuesday, Wednesday, Thursday, Friday, Saturday, and Sunday? Take into account your studying preferences and your other activities in an average week during the quarter.

Day of the Week	Hours you would study
Monday	
Tuesday	
Wednesday	
Thursday	
Friday	
Saturday	
Sunday	

5) Suppose you were offered a cash payment to sit in the library and study for a certain amount of time. You can study whatever you want (as long as it is relevant to classes you are taking) and can choose the day of the week and the time of the day when you study. Your studying can be divided up however you want but it must all take place in one day between 8am and 10pm.

What is the minimum amount of cash you would be willing to accept in exchange for studying in the library for 1 hour?

What is the minimum amount of cash you would be willing to accept in exchange for studying in the library for 2 hours?

What is the minimum amount of cash you would be willing to accept in exchange for studying in the library for 4 hours?

6) Now suppose you were offered a cash payment to sit in the library and study but instead of completing the required studying in one day, you have a week to complete the studying. The library is open every day between 8am and 10pm.

What is the minimum amount of cash you would be willing to accept in exchange for studying in the library for 10 hours in one week?

What is the minimum amount of cash you would be willing to accept in exchange for studying in the library for 20 hours in one week?

What is the minimum amount of cash you would be	willing to accept in exchange for
studying in the library for 30 hours in one week?	\$

APPENDIX E - Time-of-day studying data



### Density plots of number of subjects studying by time of day

Day of week	Morning	Afternoon	Evening	Total
Monday	3637 (8.9%)	19056 (46.8%)	18062 (44.3%)	40755
Tuesday	2047 (5.4%)	20343 (54.2%)	15171 (40.4%)	37561
Wednesday	4368 (10.8%)	20908 (51.9%)	14989 (37.2%)	40265
Thursday	1551 (5.4%)	14535 (50.2%)	12867 (44.4%)	28953
Friday	4395 (29.9%)	8368 (57.0%)	1928 (13.1%)	14691
Saturday	507 (4.2%)	9331 (77.3%)	2230 (18.5%)	12068
Sunday	619 (2.6%)	11752 (49.4%)	11416 (48.0%)	23787

Table E1: Logged minutes of study, by portion of day and day of week

The percentage of each day's study minutes is shown in parentheses. Morning was from 8:00-12:00 on weekdays, 9:30-12:00 on Saturday, and 10:20 on Sunday, afternoon was from 12:00-6:00, and evening was from 6:00-12:00 except on Saturday, when the library closed at 11:00. On occasion, the library opened at 7:30 during the week.

### **APPENDIX F** – Estimating the cost of studying

To calibrate the studying cost function, we surveyed students about how much they would need to be compensated in order to study in the library for 10, 20 or 30 hours in a week (Appendix D). Using a series of likelihood ratio tests to compare specifications of the cost function, we concluded that a simple convex cost function of the form  $c(h) = \alpha \cdot h^2$  is preferred. We estimated  $\alpha$  to be approximately 0.547 based on the regression results presented below.

Dependent Variables	(1)	(2)	(3)
Hours	7.383 (34.909)	5.01 (11.732)	-
Hours Squared	0.128 (3.356)	0.359 (0.443)	0.547*** (0.104)
Hours Cubed	0.005 (0.075)	-	-
<i>p</i> -value on Likelihood Ratio test		0.944	0.662
Observations	204	204	204
Adjusted R <sup>2</sup>	0.11	0.11	0.12
Standard errors in parentheses. *, **, and *** indicate significance at the 10%, 5%, and 1% level, respectively (two-tailed tests). We omit interaction terms, as these are not significant and do not qualitatively change the main results.			

Independent variable = Cost

The figure below shows the mean cost estimates from the questionnaire for 10, 20 and 30 hours plotted alongside the costs predicted by the chosen cost function:  $c(h) = 0.547h^2$ .



### APPENDIX G – Estimating weekly study-hour profiles in Study 1

Take the following quasi-hyperbolic objective function:

$$\min_{h_t} c(S_t, h_t) + \beta \sum_{\tau=t+1}^{t} \delta^{\tau-t} c(S_\tau, h_\tau)$$

where c(.) is the cost of studying for *h* hours in week *t*.  $\delta$  is the standard, time-consistent, exponential discount rate and  $\beta < 1$  indicates a bias for the present. We assume that agents are "sophisticated" ("naïve" behavior is not that different for the parameters we are using). *S* is the stock of remaining hours:

$$S_{t+1} = S_t - h_t$$

The student tries to minimize the cost of studying subject to the following constraint:

s.t. 
$$\sum_{t=1}^{T} h_t = S_0 = 75$$

In other words, the student must "deplete" the stock of 75 studying hours (i.e. achieve the target).

We can solve this problem numerically using the following dynamic programming equation (Harris and Laibson, 2001):

$$W(S_{t}) = \min_{h} \left\{ c(S_{t}, h) + \delta [W(S_{t+1}) - C(S_{t+1})(1 - \beta)] \right\}$$

where  $C(S_t) = c(S_t, \chi(S_t))$  and  $h_t = \chi(S_t)$  is the optimal amount of studying in week *t* with stock  $S_t$  of hours remaining given that the student knows that his future selves will choose to study  $h_{t+\tau} = \chi(S_{t+\tau})$  in all future periods.

To add in uncertainty, we assume that all future time periods are subject to a random i.i.d. shock  $z_t$ :

$$z_t = \begin{cases} 1 + \Delta \text{ with probability p} \\ \frac{1 - p(1 + \Delta)}{1 - p} \text{ with probability } 1 - p \end{cases}$$

Based on the questionnaires handed out to students, we estimate that  $\Delta=1$  and p=0.1 (i.e. there is a 10% chance that studying in the future will be twice as difficult compared to the expected cost of studying). The objective function is now:

$$\min_{h_t} c(S_t, h_t) + \beta \sum_{\tau=t+1}^T \delta^{\tau-t} c(S_\tau, z_\tau h_\tau)$$

The future uncertainty will cause the student to frontload on studying, similar to a precautionary savings model in macroeconomics (Leland, 1968). To solve the two problems, we use the

following cost function:  $c(h_t) = 0.547h_t^2$ . This cost function is derived from questionnaires handed out to students (see Appendix F).



### Simulation of study hours for different parameter values

### **APPENDIX REFERENCES**

- Harris, C. and D. Laibson (2001), "Dynamic choices of hyperbolic consumers," *Econometrica*, **69**, 935-957.
- Leland, H. (1968), "Saving and Uncertainty: The Precautionary Demand for Saving," *Quarterly Journal of Economics*, **82**, 465-473.