

HABITLAB:  
IN-THE-WILD BEHAVIOR CHANGE EXPERIMENTS AT SCALE

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DOCTOR OF PHILOSOPHY

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# Abstract

Behavior change systems help people manage their time online. However, existing productivity systems have tended to assume a one-size-fits all solution, whereas there are many factors - novelty effects, attrition, influences from other apps and devices, and differences in individual motivation - that we must take into account. That said, these effects have been researched mostly in small-scale labs studies in domains other than online behavior change, so there is a large space of opportunities for studying how these effects manifest in real-world online behavior change contexts, and how to design better behavior change systems using these insights. In this thesis we present HabitLab, an in-the-wild experimentation platform we developed for conducting behavior change experiments, as well as a set of studies we ran on the platform. HabitLab is a browser extension and mobile phone app with over 12,000 daily active, voluntary users, that users install to help them reduce time online and on their phones. It works by displaying one of 20+ interventions whenever they open an app or site they wish to spend less time on.

We use HabitLab as a large-scale experiment platform to understand behavior change. In our first set of studies, we investigate novelty effects of interventions, finding that compared to always showing the same intervention, a strategy of rotating between different interventions improves intervention effectiveness, but at the cost of increased attrition. This attrition is partly due to users being unfamiliar with rotating interventions, and improving users' mental models with a notice shown whenever a new intervention is shown is able to reduce this attrition. In our second set of studies, we investigate whether reducing time on one site or app by intensifying interventions influences time on other sites, apps, and devices. We find that on the browser, reducing time on one site reduces time spent elsewhere, but we do not observe the effect on mobile devices, and do not observe cross-device effects. In our third set of studies, we investigate users' motivation levels over time as indicated by the difficulty of interventions they select. We find that users initially overestimate how difficult of interventions they want, and their choices of difficulty progressively decline over time. Thus, we have found that online behavior change is a domain where incentives for users and researchers line up such that researchers can run large-scale in-the-wild experiments gaining ecologically valid insights about how behavioral psychology and economics theories play out in the real world, while users benefit from the more effective, scientifically informed behavior change systems that we can develop using these experiments and data.

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# Contents

<b>Abstract</b>	<b>iv</b>
<b>Acknowledgments</b>	<b>v</b>
<b>1 Introduction</b>	<b>1</b>
1.1 Thesis Overview . . . . .	3
1.2 Contributions . . . . .	4
<b>2 Related Work</b>	<b>5</b>
2.1 Behavior Change Theories . . . . .	5
2.2 Behavior Change Systems . . . . .	6
2.3 Tools Used By Behavior Change Systems . . . . .	7
2.4 Why Behavior Change Systems Fail . . . . .	8
<b>3 The HabitLab Behavior Change Experimentation Platform</b>	<b>10</b>
3.1 Introduction . . . . .	10
3.2 Mobile and Browser Versions . . . . .	14
3.3 Design of HabitLab Interventions . . . . .	14
3.4 HabitLab adoption and user demographics . . . . .	17
3.5 Design principles and tradeoffs . . . . .	19
3.6 List of Browser Interventions . . . . .	21
3.7 List of Mobile Interventions . . . . .	22
3.8 User Feedback . . . . .	23
3.8.1 Requests . . . . .	25
3.8.2 Complaints . . . . .	29
3.8.3 Positive Feedback . . . . .	32
3.9 Uninstallation Survey Responses . . . . .	32
3.10 Discussion . . . . .	32

<b>4 Rotating Online Behavior Change Interventions</b>	<b>36</b>
4.1 Introduction . . . . .	36
4.2 Related Work . . . . .	39
4.2.1 Effectiveness over time . . . . .	39
4.2.2 The impact of rotation . . . . .	39
4.2.3 Attrition . . . . .	40
4.3 Research Questions . . . . .	40
4.4 Experiment Platform: HabitLab . . . . .	41
4.5 Study 1: Field Study on the Effect of Rotating Interventions . . . . .	42
4.5.1 Participants . . . . .	42
4.5.2 Method . . . . .	42
4.5.3 Conditions . . . . .	43
4.5.4 Measures . . . . .	43
4.5.5 Method of Analysis . . . . .	44
4.5.6 Results . . . . .	45
4.6 Study 2: Longer-Term Effects of Rotation on Attrition . . . . .	48
4.6.1 Participants . . . . .	48
4.6.2 Method . . . . .	48
4.6.3 Conditions . . . . .	48
4.6.4 Measures . . . . .	49
4.6.5 Method of Analysis . . . . .	49
4.6.6 Results . . . . .	49
4.7 Study 3: Design Interventions to Reduce Attrition . . . . .	50
4.7.1 Design interventions . . . . .	54
4.7.2 Experiment Design . . . . .	54
4.7.3 Participants . . . . .	54
4.7.4 Method . . . . .	55
4.7.5 Conditions . . . . .	55
4.7.6 Measures . . . . .	55
4.7.7 Method of Analysis . . . . .	55
4.7.8 Results . . . . .	56
4.8 Discussion . . . . .	57
4.8.1 Limitations . . . . .	57
4.8.2 Design reflections on social computing and behavior change . . . . .	58
4.9 Conclusion . . . . .	59

<b>5 Do Productivity Interventions Save Time or Just Redistribute It?</b>	<b>60</b>
5.1 Introduction . . . . .	60
5.2 Related Work . . . . .	63
5.2.1 Multitasking and Cyberslacking . . . . .	63
5.2.2 Distribution Of Unproductive Time . . . . .	63
5.3 Research Questions . . . . .	64
5.4 Experiment Platform: HabitLab . . . . .	65
5.5 Study: Redistribution of Time Within and Across Devices . . . . .	65
5.5.1 Participants . . . . .	65
5.5.2 Method . . . . .	66
5.5.3 Intensity . . . . .	66
5.5.4 Time Redistribution . . . . .	67
5.6 Results . . . . .	68
5.6.1 Are interventions effective? . . . . .	68
5.6.2 Is time spent on goals reduced when there is higher intensity? . . . . .	69
5.6.3 What is the effect of increasing intensity on other, non-goal apps and sites? . . . . .	70
5.6.4 Is time redistributed between devices? . . . . .	71
5.6.5 Destination tracking . . . . .	72
5.7 Limitations . . . . .	72
5.8 Discussion . . . . .	75
5.9 Conclusion . . . . .	76
<b>6 Discussion</b>	<b>77</b>
6.1 Design Principles for In-the-wild Experimentation . . . . .	77
6.2 Visions for the Future of Behavior Change Systems . . . . .	78
6.3 Limitations . . . . .	79
<b>7 Conclusion</b>	<b>80</b>
<b>A Replication of Chapter 4, Study 1 Findings using Session Level Measurements</b>	<b>81</b>
A.1 Effectiveness of interventions over time . . . . .	81
<b>B List of Interventions used in Chapter 4 Studies</b>	<b>83</b>
<b>C List of Interventions used in Chapter 5 Studies</b>	<b>85</b>
C.1 List of Browser Interventions . . . . .	85
C.2 List of Mobile Interventions . . . . .	87
<b>D User-Contributed Intervention Ideas</b>	<b>88</b>

<b>E User Feedback on GitHub Issues</b>	<b>107</b>
<b>F Responses to Uninstall Survey</b>	<b>128</b>
<b>Bibliography</b>	<b>182</b>

# List of Tables

3.1	A subset of the interventions for Facebook, categorized according to persuasion strategy and theory. Interventions that are enabled by default are marked <i>default</i> , interventions that are available for all sites are marked <i>generic</i> . . . . .	17
4.1	Within the static condition, interventions decline in effectiveness. Longer visit lengths increase with the number of days seeing the same static intervention. . . . .	45
4.2	Daily time spent on sites in the static and rotation conditions. Users spend less time per day on sites in the rotation condition. . . . .	46
4.3	A Cox proportional hazards analysis suggests that the rotation condition substantially increases the hazard of attrition. Coefficients are log hazard ratio, so positive values indicate increased hazard and negative values indicate decreased hazard. . . . .	46
4.4	A Cox proportional hazards analysis over a longer period suggests that rotating with more interventions increases the hazard of attrition. . . . .	50
4.5	A Cox proportional hazards analysis suggests that the informational intervention that corrected users' mental models was successful in reducing attrition due to rotation. Coefficients are log hazard ratio. . . . .	57
5.1	Data Summary. Note that the duration of 132 days are users who kept it installed the longest, but as users can freely install/uninstall we do not have 132 days of data on all users. . . .	66
5.2	Browser: Frequent interventions for a goal site cause a reduction of time spent on the site. .	68
5.3	Mobile: Frequent interventions for a goal app cause a reduction of time spent on the app. .	69
5.4	Browser: Increasing intensity results in a reduction of time spent each day on all goal domains	69
5.5	Mobile: Increasing intensity results in a reduction of time spent each day on all goal apps .	70
5.6	Browser: Increasing intensity results in a reduction of time spent each day on non-goal sites	70
5.7	Mobile: Increasing intensity has no significant effect of time spent on non-goal apps. . . .	71
5.8	Mobile: Varying intervention intensity has no effect on total time spent on browser goal sites	71
5.9	Browser: Varying intervention intensity has no effect on total time spent on mobile goal apps	72

A.1 Within the static condition, interventions decline in effectiveness, with longer visit lengths with increasing larger number of days since it was first observed. . . . .	82
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# List of Figures

1.1	Examples of research questions that can be studied with the HabitLab system and the general space of questions they occupy. Research questions we study in this thesis are shown in green.	2
3.1	Screenshots from the mobile version of HabitLab. Left: The goal selection screen, where users choose which apps to spend less time on. Right: An example intervention, which shows the visit count when a user opens a goal app.	11
3.2	The goal selection screen, where users choose which sites to spend less time on (browser version).	12
3.3	An example intervention, which asks a user to write their objective for visiting a site (browser version).	13
3.4	HabitLab's homepage describes the browser extension and mobile application. Users adopt it to try out a large number of different possible interventions, called nudges.	14
3.5	During onboarding, users choose which sites they want to spend less time on.	15
3.6	Users are presented with the interventions they will see on each site.	16
3.7	Examples of interventions available for reducing time on Facebook. From left to right, top to bottom: a timer injected into the news feed; a page before opening Facebook requiring that the user wait a few seconds before visiting; a countdown timer that automatically closes the tab after time elapses; an opt-in required to show the news feed; an interstitial page before opening Facebook with a quote; an interstitial page before opening Facebook that requires the user set a time limit for how long they will spend this session.	18
3.8	Ages of HabitLab users. 25-35 is the most-represented demographic.	18
3.9	Map of countries representing HabitLab users. North America, Europe, and Asia are all well-represented.	19
3.10	Top 10 countries using HabitLab. Users from the US account for 30% of our userbase.	20
3.11	Languages that HabitLab users set as their preferred language to show webpages in. English is the preferred language of half of our users.	20
3.12	Our interface for submitting feedback within HabitLab.	24
3.13	Our interface for letting users submit new intervention ideas and vote for existing ones.	26

3.14	The survey shown to users if they uninstalled. . . . .	33
3.15	Reasons users selected for uninstalling, in responses to our uninstallation survey. . . . .	33
4.1	Example order in which a user might see conditions. Each circle represents a day – on black days the user is in the “static” condition, white is the “rotation” condition. The order of blocks is randomized; here, this participant is seeing blocks in order 1, then 3, then 5, then 7 (omitted in the figure). . . . .	43
4.2	Rotating interventions increases attrition among users. . . . .	47
4.3	Including all interventions resulted in significantly more attrition than just one intervention. .	49
4.4	Users who change interventions during the first 5 minutes will be significantly less likely to attrition later. . . . .	51
4.5	Mental model interface: each time the user sees a new intervention, HabitLab names it and explains about rotation. . . . .	55
4.6	User control interface: in addition to the mental model information, HabitLab gives users a direct interface to disable the new intervention. . . . .	55
4.7	Reminding users about how the rotations worked every time a new intervention was introduced significantly reduced attrition rates. . . . .	56
5.1	When interventions reduce time on a targeted goal such as Facebook, the time saved may (left) be isolated from effects on other goals, (center) be redistributed to other goals, or (right) decrease time spent on other goals. . . . .	61
5.2	When interventions reduce time on a targeted device e.g. a browser, the time saved may (left) be isolated from effects on other devices, (center) be redistributed to other devices, or (right) decrease time spent on other devices. . . . .	61
5.3	The top 10 goal apps with the most number of sessions on mobile are on the left. On the right is the distribution of where a user ends up immediately after. . . . .	73
5.4	The top 10 goal apps with the most number of sessions on the browser are on the left. On the right is the distribution of where a user ends up immediately after. . . . .	74

# Chapter 1

## Introduction

We wish to spend our time more productively, but we sink hours into social media; we wish to learn new languages, but we get too busy to practice; we wish to be more healthy, but we do not maintain our exercise routines [34]. Inspired by situations like these, *behavior change* systems help people build new habits and retain them [35, 59, 81, 85]. Behavior change systems draw on theories of persuasion and influence [56, 31] to introduce *interventions*: interaction designs that variously inform, nudge, and encourage people to engage in behaviors more in line with their goals.

There are large numbers of users who wish to achieve behavior change goals, and a large design space for interventions. Thus, there is a natural opportunity to explore the design space of interventions and find what interventions works best, by testing them out with users. However, the existing ecosystem of behavior change tools does not make full use of this resource.

Behavior change tools, both those tailored towards commercial mass-market adoption by end users as well as research systems, have tended to employ a one-size-fits-all approach, implementing only a single behavior change intervention and giving the same experience to all users. Users can choose and select between different behavior change apps and extensions to find what they believe works well for them. However, because different apps are developed by different companies which do not share data, they cannot compare the interventions. They thus miss out on a rich opportunity for improving the behavior change systems via experimentation.

Research studies on behavior change, in contrast, have tended to compare only a small number of interventions, with small numbers of paid participants. They thus miss out on the ecological validity, scale, and statistical power that systems targeting the mass market of end users can enjoy.

Our key insight was that we can find an alignment the goals of behavior change researchers and end users, by building a behavior change tool targeted towards the mass market that also runs useful experiments. End users benefit by being able to use a high-quality behavior change tool where the design choices are experimentally tested and validated. Researchers benefit by being able to run ecologically valid behavior change experiments at scale.

Examples of research questions we can potentially study using HabitLab			
Question type	Questions related to goals of users	Questions related to choice of intervention	Questions related to outcomes of interventions
Observed state	What goals do users have?	What interventions do users choose?	What interventions are effective?
Changes	How and why do user goals change over time?	How and why do intervention choices change over time?	How and why does intervention effectiveness change over time?
Measures	How do users' stated goals differ from their actual goals?	How do users' stated intervention preferences differ from their actual choices?	How does effectiveness measured for an intervention differ from overall effect?

Figure 1.1: Examples of research questions that can be studied with the HabitLab system and the general space of questions they occupy. Research questions we study in this thesis are shown in green.

While we believe that many domains can potentially benefit from in-the-wild behavior change experimentation, we believe that online behavior change is particularly well-suited. As interventions can be distributed as software that requires only a click to install, this enables us to recruit a wide range of participants worldwide for free. Additionally, as our computers and phones can display arbitrary interactive content, this paradigm allows us to experiment with a limitless number of different interventions, and change interventions at any time. Finally, as device usage can be precisely monitored down to the level of which webpage or app was open each second, we can easily measure the effectiveness of interventions and adapt them accordingly.

As a result, we built HabitLab, an in-the-wild experimentation platform for helping users reduce their time online and on their phones. HabitLab is implemented as both a Chrome extension and an Android app, and is currently used by over 12,000 daily active users. Users select sites and apps they wish to spend less time on, and HabitLab deploys a variety of interventions to help them achieve their goals. The platform enables us to run a number of A/B tests comparing interventions and aspects of behavior change systems.

There is a rich set of studies we can run with a tool such as HabitLab. The general paradigm is that users specify goals, interventions are deployed to help them achieve those goals, and we measure the outcomes of the interventions. At a high level, we can thus categorize the space of possible research studies that a platform such as HabitLab can conduct as below. A more detailed classification is shown in Figure 1.1.

1. Studies analyzing users' goals
2. Studies analyzing the choice of interventions to help users achieve those goals
3. Studies analyzing the outcomes of interventions

We believe that the HabitLab system is uniquely well-suited for investigating the latter two classes of problems within behavior change. Specifically, it allows us to quickly test a variety of different choices of interventions – as we can make arbitrary changes to websites, or show arbitrary overlays over the phone screen. Additionally, it also enables us to precisely measure the outcomes of the interventions – specifically, we can directly observe the amount of time users spend on sites.

In addition to demonstrating the technical feasibility of such an in-the-wild experimentation system by building it and using it for studies, we demonstrate via HabitLab’s large install base that it is possible to align user incentives with those of researchers within this behavior change domain. Many other studies rely on having to pay or force participants to use the system, as the primary beneficiary of the research is the researcher – but with HabitLab, users benefit from the research, as the insights we gain from user behavior data allows the system to better help users achieve their goals.

## 1.1 Thesis Overview

In this thesis we will first present the HabitLab system. Then we will describe a pair of studies conducted on the HabitLab system, which both study secondary effects that influence the intervention outcomes that we observe.

### **Chapter 3: The HabitLab System**

We first describe the HabitLab system itself, the design process, the interventions, how we were able to amass a userbase of 12,000 daily active users via an in-the-wild deployment, and the demographics of those users.

### **Chapter 4: Intervention Rotation Study**

The first study we present asks whether the effectiveness of interventions decline over time. Prior literature suggests this is a possibility, as engagement-boosting novelty effects have been attested in numerous domains. We find that an intervention that is repeatedly presented does indeed decline in effectiveness over time, and that a strategy of rotating between different interventions helps boost effectiveness. This boost comes at the cost of increased attrition, which we find mostly to be due to incorrect mental models, as users are unaccustomed to interventions changing. A simple design helping explain the intervention rotation to users and give them a sense of control significantly reduces this resulting attrition.

### **Chapter 5: Time Redistribution Study**

The second study concerns itself with whether intervention outcomes are actually what we expect them to be looking at just the time spent on the targeted site or app. Prior literature suggests that willpower is limited, hence we may expect that reducing time on one app, site, or device may increase time spent on

others. Other literature suggests that procrastination begets more procrastination by trapping us into a habit loop, hence we may expect that reducing time on one app, site, or device may reduce time spent on others. We find that in the case of site usage on browsers, reducing time on one site results in a reduction of time spent on others. In the case of app usage on mobile, we observe that reducing time on one app does not affect time on others. Likewise, in the case of devices, reducing time on one device does not affect time on others.

## 1.2 Contributions

The contributions of this thesis are:

1. HabitLab, a system for conducting in-the-wild behavior change experiments online. It has been widely adopted with a 12,000 user install base across the browser and mobile platforms, showing that in-the-wild behavior change systems can be used to conduct large-scale experiments.
2. A set of design principles we have developed for in-the-wild experimentation platforms, which help researchers navigate conflicts that may emerge between the needs of end users and the scientific needs of researchers.
3. A set of studies conducted on HabitLab showing that static interventions decline in effectiveness over time, and that rotating interventions can improve their effectiveness.
4. A set of studies conducted on HabitLab showing that interventions can sometimes have beneficial secondary effects, reducing time spent on non-targeted sites.

These studies show that our paradigm of in-the-wild experimentation, as realized in the domain of online behavior change via the HabitLab, can work to find novel insights about behavior change systems. We hope this work can help designers build better systems for online behavior change, and promote analogous in-the-wild experimentation in other behavior change domains.

# **Chapter 2**

## **Related Work**

In this chapter we will cover work in sociotechnical systems, psychology, and behavioral economics related to behavior change and persuasive technology. We will begin with theoretical frameworks and taxonomies about behavior change, discuss examples of behavior change systems, discuss various tools which behavior change systems make use of, and finally explore the causes of attrition and why behavior change systems fail.

### **2.1 Behavior Change Theories**

#### **Theoretical Frameworks for Behavior Change**

The field of persuasive technology studies how technology can be used to influence behavior [56]. There are a number of theoretical frameworks describing behavior change systems. B-MAT is a popular framework of behavioral change [56], which demonstrates that systems can focus on three elements—motivation, ability, and a trigger (a call to action)—to produce behavior change. The habit loop is another framework for building habits [50], stating that systems can build habits through an iterated process of displaying a trigger, prompting the user to take an action, giving out a reward, and helping the user to invest in the system.

#### **Taxonomies of Behavior Change**

A number of taxonomies characterize the design space of interventions, both general [107, 108, 2, 45] and domain-specific [62, 166]. Michie’s behavior change taxonomy lists 93 techniques for behavior change, clustered according to the cognitive phenomenon they target [107]. Systems have investigated effects of these techniques individually, such as using “cheating” to support lapse management [5], using different framings to present results [85], or setting goals and plans [6].

## 2.2 Behavior Change Systems

Persuasive technology systems have been successful in promoting behaviors such as sustainable resource usage [59], fitness [35], sleep [81, 27], healthy eating [115, 47], stress management [3, 140], smoking cessation [118], and productivity [168, 85].

They can operate on many different platforms, such as the web or mobile devices. Web-based systems promote a behavior change goals including classroom engagement [12, 11], psychology therapy [16] and healthy habits [38, 100]. In parallel, a number of studies focused on mobile-based interventions [119, 135, 54, 167, 128]. For instance, MyBehavior, a mobile phone app, was built to track physical activities of the users and to provide personalized suggestions that are tailored to the users' historical behavioral data [128]. Similarly, PopTherapy is a mobile phone app that studied micro-interventions for coping with stress [119].

### Sociotechnical Systems for Behavior Change

People use a variety of sociotechnical systems to support behavior change, including forums [52, 26], social sharing [124, 30, 123, 87], personal informatics [94, 29], and self-experimentation [80]. People use behavior change forums to gain social support [66] – meeting social needs such as approval and esteem [78]. They do so by providing users with information and advice [66], and establishing norms [26]. They also facilitate social comparisons [42] which influence behaviors, as social comparison theory states that users seek to bring their behaviors in line with norms [53]. Communities also help users find others with similar experiences [67] who can help them through the process of recovering and adapting to changes [112]. Social sharing [124, 133] works by helping users receive support through social interactions, and encouraging accountability [48]. Personal informatics support behavior change through stages of preparation, collection, integration, reflection, and action [94]. The theory of lived informatics [49] adds additional stages where users choose tracking tools, and alternate between lapsing and resuming their tracking behaviors.

### Online Behavior Change

One major topic inspiring our work is users' desires to curb or control their time spent on social media sites. People pressure themselves to, and often do, make efforts to reduce their time spent on social media sites such as Facebook and Twitter [147, 141]. Yet this is difficult because users turn to social media to address their need to belong, the need for self-presentation, the need for self-esteem [110], the need for entertainment and gratification [126], and self-affirmation [158]. Whether social media use improves well-being is a complex question depending on the nature of the engagement [162, 102, 96, 84, 109, 138, 154], but thanks to instant gratification and sites' use of gamification [28, 170, 68] and behavior design techniques [56, 50] to drive engagement, users keep coming back to the point that some consider it an addiction [13, 137, 155, 161].

### Comparison of HabitLab to Existing Systems

There exist several other behavior change systems dedicated to helping users reduce time online. PopHistory [24] is a browser-based system which shows a visualization of where users' time is spent. TimeAware [85] is another browser-based time tracker, where the researchers investigated effects of different ways of framing how time was spent. A browser extension by Agapie et al both tracks time and deploys an interstitial page when visiting a goal site, where they studied the effects of providing users with "cheat points" to reduce user attrition [5]. Commercial online behavior change systems include RescueTime [132] and Forest [142], which are time tracking visualizations for browsers and mobile devices, as well as StayFocused [160], Self-Control [153], and LeechBlock [71], which includes a site blocking intervention. In comparison to these, HabitLab is unique in that it includes a variety of different interventions that we can compare via experimentation, as opposed to only a single one. Some behavior change systems in other domains, such as PopTherapy [119], also include various interventions which they compare using a multi-armed bandit algorithm. Compared to the stress domain investigated in PopTherapy, the HabitLab system has the advantage of being in a domain where we can measure efficacy directly instead of relying on self-reported data. We also have a much longer deployment and userbase, which enabled us to run a number of studies investigating secondary effects of behavior change interventions.

## 2.3 Tools Used By Behavior Change Systems

Researchers in behavioral economics and related fields have developed a number of tools which are useful for building behavior change systems. Gamification, which introduces triggers, investment, rewards, and game-like elements to motivate behavior change, and personalization are a pair of tools which are often used to boost the effectiveness of behavior change systems. Another one of these tools is *choice architectures*, or ways to structure choices to influence behaviors.

### Gamification

Much previous work has focused on gamification as an approach to design behavior change systems [44]. Gamification has been shown to have positive effects on engagement and outcomes in behavior-change contexts such as promoting healthy habits [38, 100] and improving educational engagement [11, 12], though effectiveness varies depending on the context and design [60].

### Personalization

A recent trend in behavior change systems has been the concept of personalizing interventions. Such systems explore several possible strategies using techniques such as multi-armed bandits to find the intervention that is most effective for the user [119, 127]. For example, PopTherapy demonstrated personalized messaging could be found through such techniques [119]. Likewise, HeartSteps conducted tens or hundreds

of micro-randomized trials on users [43].

### Choice architectures

The field of behavioral economics has developed a number of theoretical frameworks for how to present choices to influence people's choices, known as choice architectures [156, 157, 75]. One of the best-known choice architectures is defaults – the choice made if the user does not make an active choice – which work by exploiting the status quo bias [139]. Defaults have been found to be effective in numerous behavior change contexts, including increasing organ donations [73], encouraging saving for retirement [37, 101], and influencing choice of insurance plans [74]. Other examples of effective, widely deployed choice architectures include limiting the number of choices [37, 86], sorting choices [99], grouping choices [57], and simplifying choice attributes to be more easily interpretable [122, 148]. Some choice architectures are designed explicitly for interactive behavior change contexts where users repeatedly make choices, such as Enhanced Active Choice, which provides users with choices while attempting to steer them towards the desired one [82].

A number of choice architectures have been developed to combat our bias towards myopic choices, aversion to uncertainty, and lead us to choices that have better long-term outcomes [75]. If the user is choosing between short-term benefits and longer-term benefits, one strategy is to make the long-term outcomes of the choice more salient in the short term [165, 149]. In cases where there are a large number of uncertain outcomes and we wish to encourage satisficing – that is, choosing an acceptable option sooner rather than waiting for a hypothetical future optimal choice – we can focus attention on the second-best outcomes [145]. This strategy of satisficing can both lead users to decide faster, and leave them happier with their choices [69]. We can also combat procrastination resulting from users' tendency to be overly optimistic about future opportunity windows by explicitly enforcing limited opportunity windows [116].

## 2.4 Why Behavior Change Systems Fail

While users are often initially highly motivated to use a behavior change system, many of them will eventually attrition, or stop using the system. Attrition and other failures of behavior change systems may be due to a number of factors, including declining user motivation, unrealistic initial expectations, and conflicts between short-term rewards and long-term goals. Researchers in the field of behavioral economics have conducted numerous studies about users' motivation levels, how they change with time, and whether users can accurately predict their future motivation levels.

### Attrition

Attrition is a major challenge in behavior change systems. Attrition [51], also known as dropout, occurs when participants stop participating, leave, or uninstall the system. A metastudy of eHealth interventions found that an attrition rate around 99% over a 12-week period is normal [51]. Likewise, the number of users in a stress-coping mobile application declined in a steady rate through the study [119]. Persuasive systems

built for weight control and therapy have shown substantial attrition rates in longitudinal studies [21, 119], and prior work in CSCW has sought to help reduce attrition rates through techniques drawn from dieting and addiction research [5].

### **Optimism vs Unrealistic Optimism**

Are users initially too optimistic and have unrealistic expectations? Or are they able to accurately judge their levels of self-control? Does optimism have positive or negative results in behavior-change contexts?

Optimism – a generally positive and hopeful attitude towards their future abilities to change – is generally considered to be beneficial in behavior change, and is considered to predict positive outcomes [41]. Behavior-change literature makes a distinction between this optimism and unrealistic optimism – where the subject underestimates their susceptibility to health problems – which is considered to be detrimental towards behavior change outcomes [41].

In dieting contexts, users tend to overestimate their self-control abilities and have unrealistic expectations of their ability to lose weight [150]. The level of participants' optimism – how much they underestimate their ability to – can be predicted based on survey questions related to their self-identity and theory of planned behavior. However, users differ in the levels to which they are optimistic [41].

### **Short-Term Myopia vs Long-Term Goals**

Users often make short-term choices that conflict with their long-term goals, which often manifest themselves as inability to delay gratification, lack of self-control, procrastination, and addiction [98, 146]. These myopic choices can be attributed to a number of factors – firstly, short-term benefits are more immediate and salient than long-term losses, leading us to discount future outcomes [7, 98, 149]. The degree to which users place decreasing value on future outcomes has been studied in various economics experiments on intertemporal discounting [165], and models such as hyperbolic discounting and subadditive discounting have been proposed to describe this bias in mathematical terms [8, 136, 130].

Additionally, we are often certain of short-term benefits, while long-term effects such as life expectancy are less certain, so we discount the uncertain, long-term outcomes [120], or end up considering only a desireable subset of possible outcomes [145, 88]. Optimism can also play a role in myopic choices, as we are often overly optimistic that we will not suffer from possible negative long-term consequences [77, 169].

## Chapter 3

# The HabitLab Behavior Change Experimentation Platform

### 3.1 Introduction

Studying behavior change requires in-the-wild intervention and observation [35]. Inspired by previous CSCW tools for naturalistic data collection [131], we developed HabitLab [90], an open-source<sup>1</sup> platform, as a living laboratory to help us understand online behavior change and as a platform to explore novel behavior change designs (Figure 3.4).

The contribution of the HabitLab system is a behavior experimentation platform that enables researchers to run a number of studies at scale to develop novel, field-tested insights about behavior change – which we will discuss examples of in later chapters – while providing benefits to end users – as evidenced by our large active, voluntary userbase. We show that while conflicts between user preferences and research goals often exist, we can design our experiments to navigate this tradeoff and run scientifically sound experiments without sacrificing user satisfaction, retention, and growth.

HabitLab is a Chrome browser extension and Android application that contains a variety of productivity interventions. It aims to help users reduce their time spent online on web pages that the user specifies (e.g., Facebook, Twitter, and Reddit). The system is pitched to end users as a tool that explores various different interventions (referred to as “nudges”) to help them reduce their time on sites.

Both versions follow the structure of allowing users to choose what they wish to spend less time on (setting goals), and deploying interventions to meet those goals. On the Chrome version, users choose sites to spend less time on (goal sites – for example, `facebook.com`), as shown in Figure 3.2. On Android, users choose particular apps to spend less time on (goal apps – for example, the Facebook Android app), as shown in Figure 3.1. Interventions are deployed when users visit a goal site on Chrome (Figure 3.3), and when users

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<sup>1</sup>HabitLab is available at <http://habitlab.github.io>.

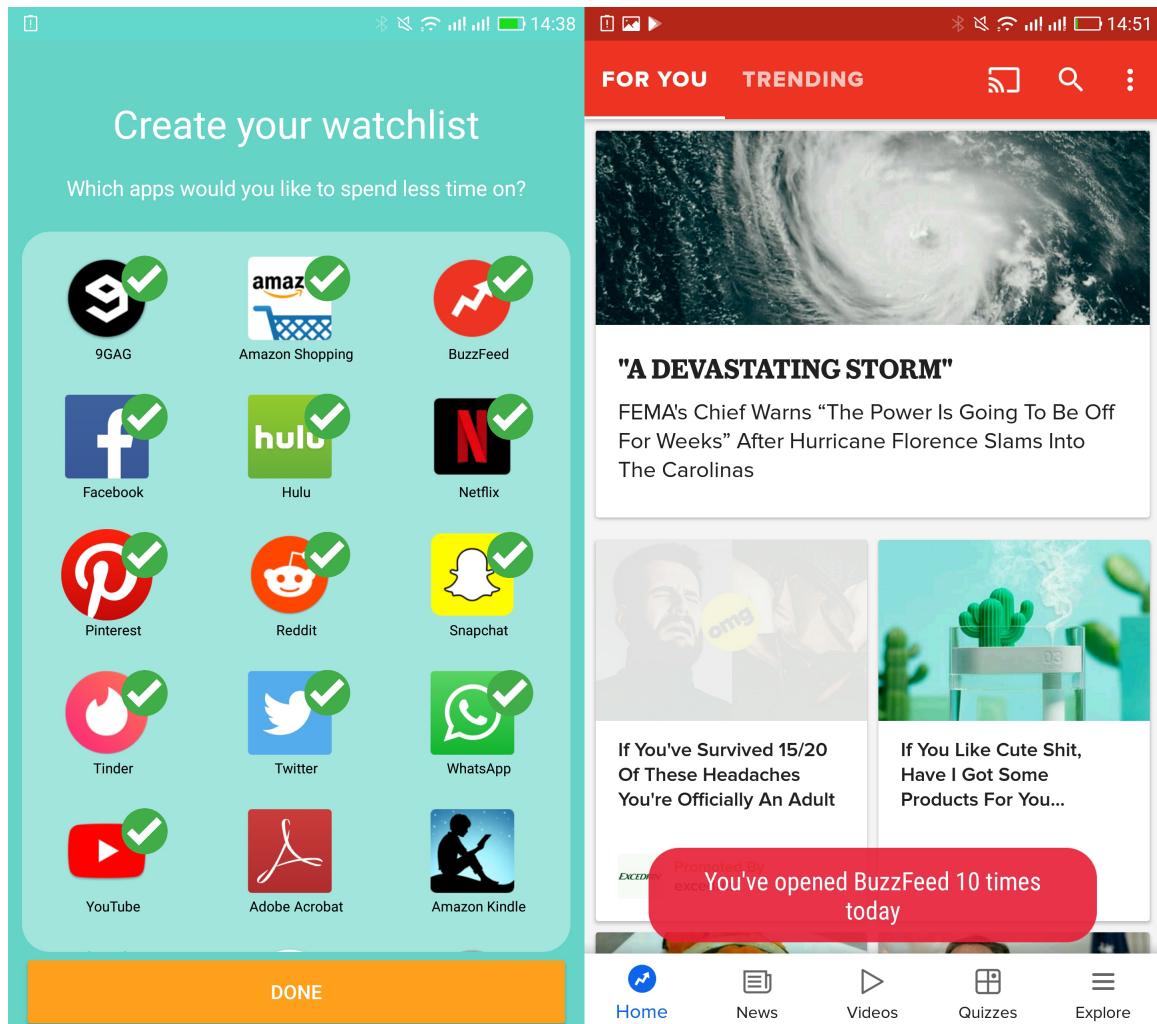


Figure 3.1: Screenshots from the mobile version of HabitLab.

Left: The goal selection screen, where users choose which apps to spend less time on.

Right: An example intervention, which shows the visit count when a user opens a goal app.

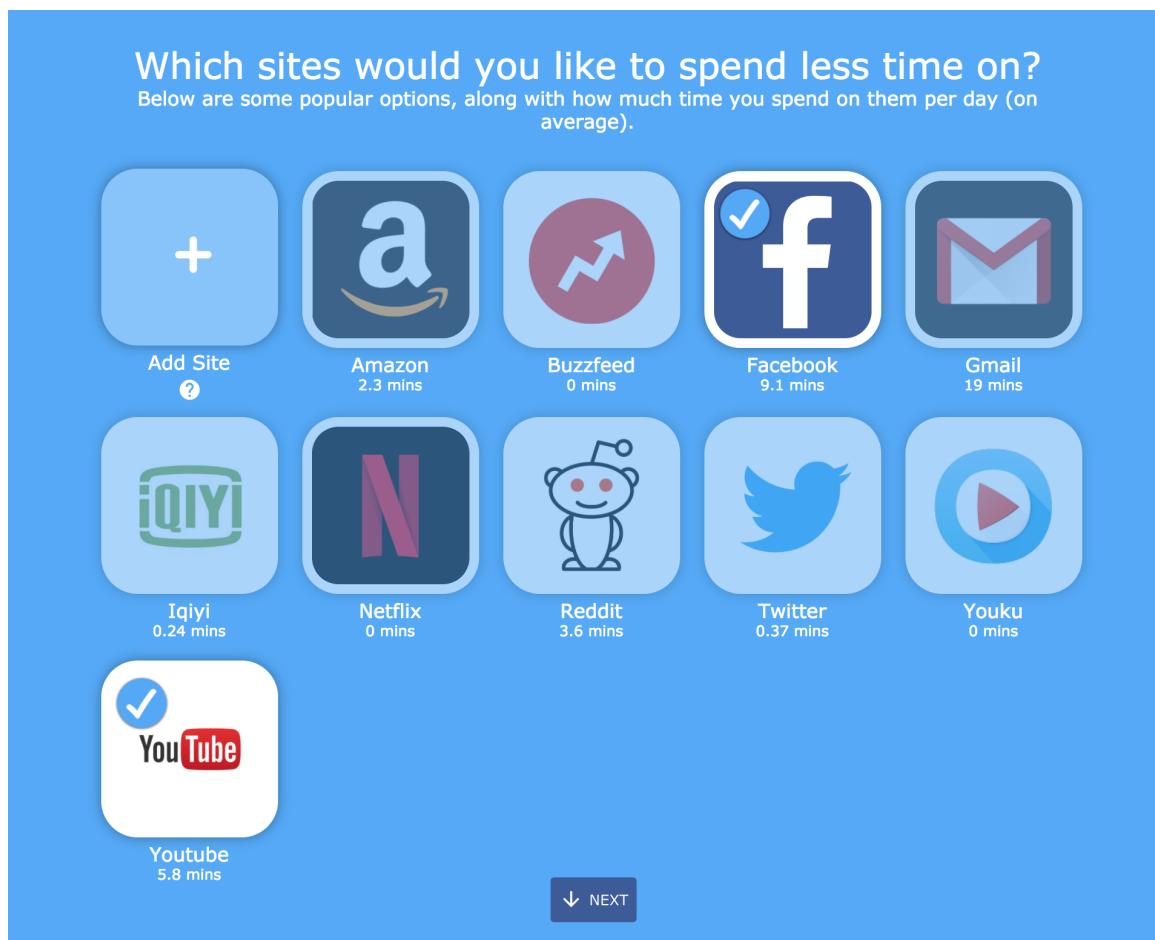


Figure 3.2: The goal selection screen, where users choose which sites to spend less time on (browser version).

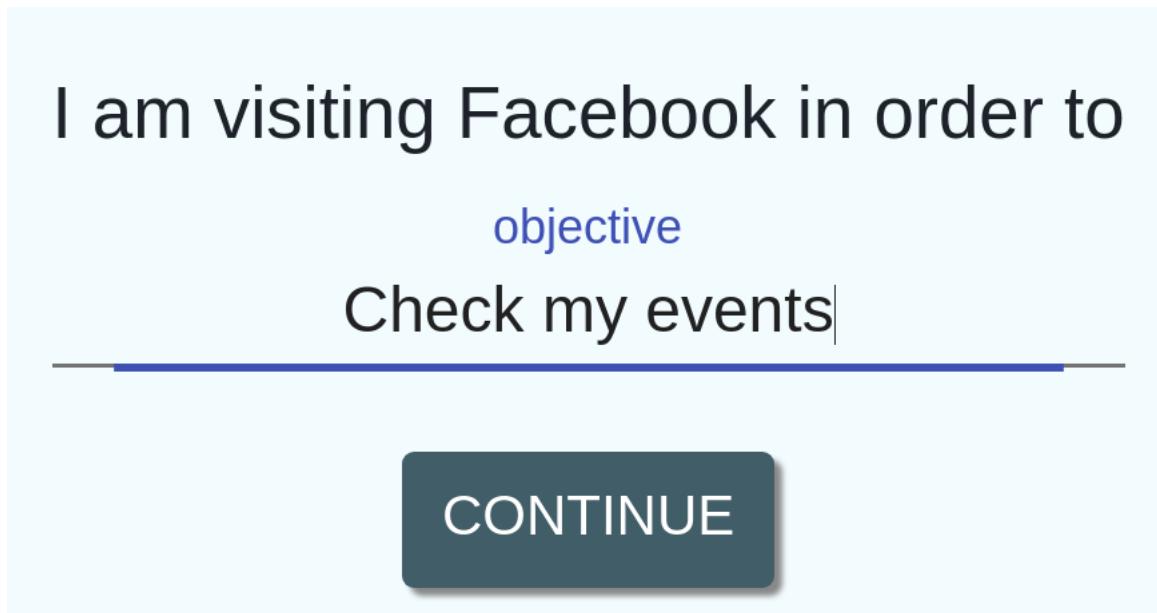


Figure 3.3: An example intervention, which asks a user to write their objective for visiting a site (browser version).

open a goal app on Android, as shown in Figure 3.1.

Users install the extension, and go through an onboarding process where they select sites and apps they wish to reduce their time on (Figure 3.5). There are predefined options—Facebook and YouTube are selected by default, as they were the most commonly used—but users can also add any custom site. Custom sites are suggested via an analysis of the user’s browsing history. The system explains to users that they will be shown a variety of interventions (Figure 3.6), a form of self-experimentation [80], to help them reduce time on that site. These interventions are typically targeted to each site, for example a news feed blocker for Facebook or a related video hider for YouTube. However, some interventions such as a stopwatch timer can be added to any custom site. Users can preview the interventions for the sites they select, and enable or disable each intervention if desired. Users can later enable or disable interventions and sites through a settings page.

HabitLab emphasizes to users the availability of multiple interventions and that it may show users different interventions each time they load a page. This emphasis is made clear on the HabitLab website, Chrome store listing, and through features in the dashboard such as visualizing the relative effectiveness of different interventions. HabitLab implements a multi-armed bandit algorithm to explore and find the interventions that are most effective for each user, optimizing for minimizing time spent on a site.



Figure 3.4: HabitLab’s homepage describes the browser extension and mobile application. Users adopt it to try out a large number of different possible interventions, called nudges.

## 3.2 Mobile and Browser Versions

The Chrome extension and Android app differ in some minor details. They support different sets of goals: users select apps to reduce time on in the Android version, whereas users choose sites to reduce time on in the Chrome version. Additionally, the specific set of interventions available differs between the platforms to fit the design languages of the browser and the mobile phone. The Chrome version has certain interventions which are site-specific – such as a news feed remover that is specific to Facebook. However, because Android does not allow applications to edit each other’s view trees, the Android version’s interventions are all glass pane overlays, and thus are general and can be used on any app. The concept of a session is different on the platforms: in the Chrome version, a session is time on a site until that tab is either closed or the user goes to a different domain. Time measured is active time – so if the tab is not focused, or if there is no keyboard or mouse activity for over a minute, the timer is temporarily paused. However, on Android, because there is no concept of a tab, the measurement of a session is different. There, a session is considered the duration over which an app is opened and focused. Closing the app, switching to a different app, or turning off the phone will end the current session.

## 3.3 Design of HabitLab Interventions

HabitLab can track time and deploy interventions on all sites, but some interventions are tailored towards specific sites. There are 27 interventions total: seven generic interventions that can be used on all sites, five interventions designed specifically for Facebook, and additional ones designed specifically for YouTube,

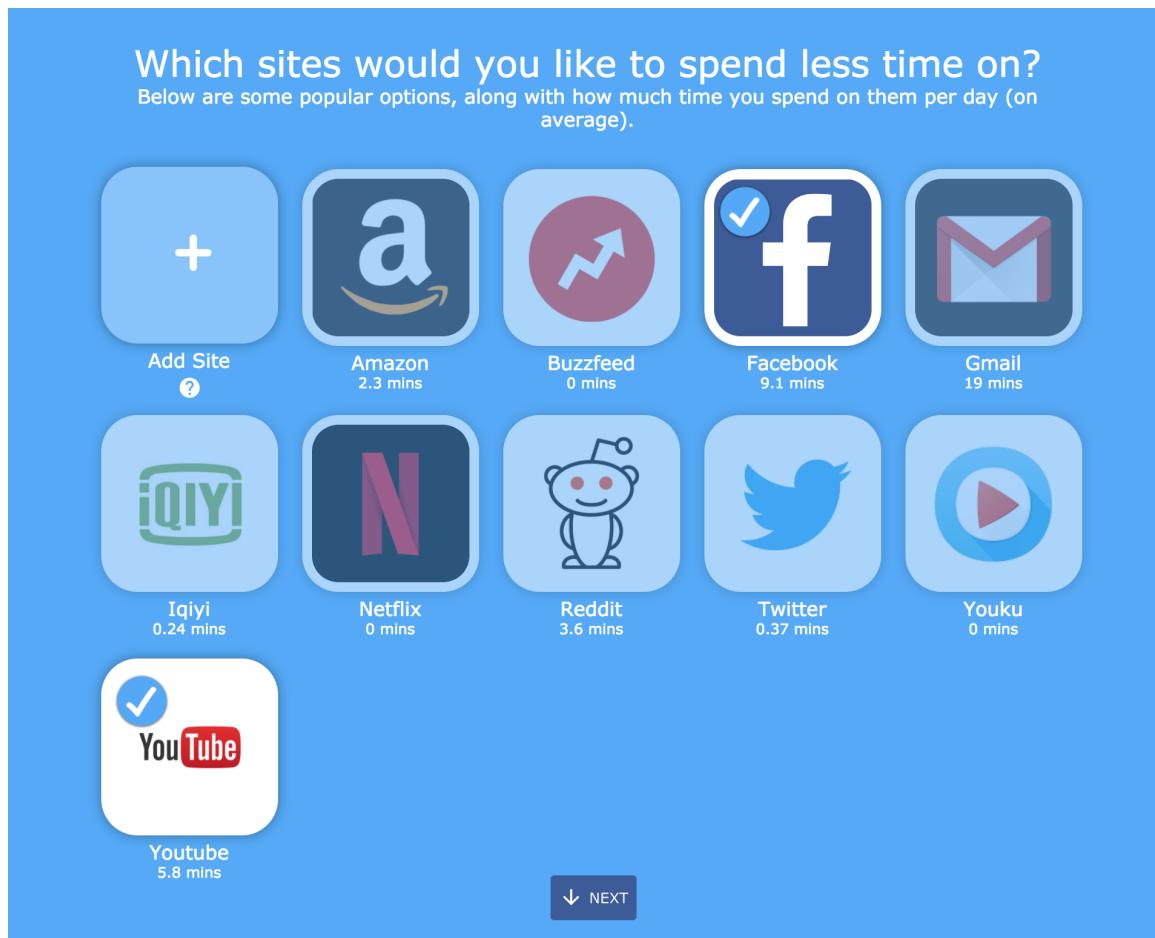


Figure 3.5: During onboarding, users choose which sites they want to spend less time on.

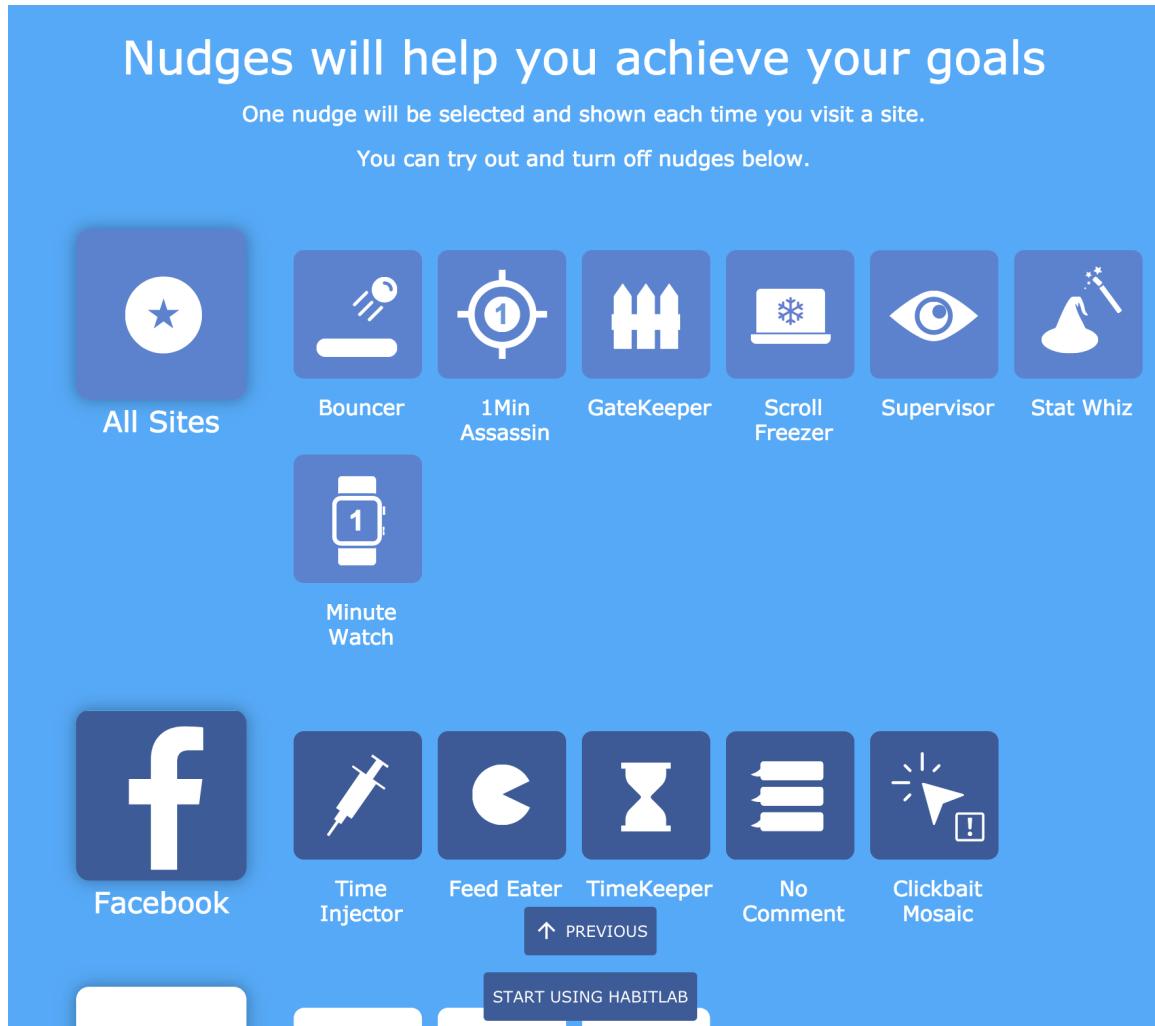


Figure 3.6: Users are presented with the interventions they will see on each site.

Strategy	Theory	Intervention
Commitment	Self-consistency theory [10, 31, 144]	Ask the user to set a goal for the length of time they will stay on the site (generic)
Enforce default limits	Status quo bias [139]	Automatically close tab after 60 seconds unless the user clicks a button to ask for more time (generic)
Reduce social incentives	Social proof [143, 31]	Hide Facebook comments by default (default)
Delaying Rewards	Operant conditioning [17]	Make the user wait 10 seconds before visiting Facebook (generic)
Removing Rewards	Operant conditioning [17]	Hide the news feed (default)
Inform the user	Theory of reasoned action [9]	Show a counter at the top of the page of how long user has been on Facebook today (default, generic)

Table 3.1: A subset of the interventions for Facebook, categorized according to persuasion strategy and theory. Interventions that are enabled by default are marked *default*, interventions that are available for all sites are marked *generic*.

Reddit, Twitter, Netflix, Gmail, Amazon, iQiyi, and Youku.

Interventions are designed drawing on theories of behavior change—for example, goal setting theory [97], persuasion [31, 56, 2], and gamification [44]. A sample of the interventions available for Facebook, categorized according to underlying strategies and theories, are shown in Table 3.1. Screenshots of some Facebook interventions are shown in Figure 3.7. Descriptions of the interventions on the Chrome and Android versions can be found at the end of this chapter.

Not all interventions are enabled by default—this is because some of them have higher attrition rates than others. Non-default interventions can be previewed and enabled by users during onboarding and on the settings page. The interventions enabled by default were the ones we found to have low attrition rates during pilot deployments—we chose this strategy to ensure user retention and growth, which is a prerequisite for gathering data in an in-the-wild experiment setting.

### 3.4 HabitLab adoption and user demographics

As of writing, the browser version of HabitLab has over 12,000 daily active users, and the Android version has over 500 daily active users.

Demographics according to Google Analytics indicate that our users are 81% male, with the most commonly represented age group being 25-34, as shown in 3.8.

Our userbase represents a diverse set of countries and languages – users represent 151 countries as shown in 3.9. The top 10 countries are shown in 3.10 – the US is the most-represented country, representing 30% of the userbase.

Half of our userbase uses English as their preferred language for displaying webpages, as indicated in 3.11. Volunteers have translated HabitLab into 13 languages (Arabic, Chinese, Czech, Dutch, French,

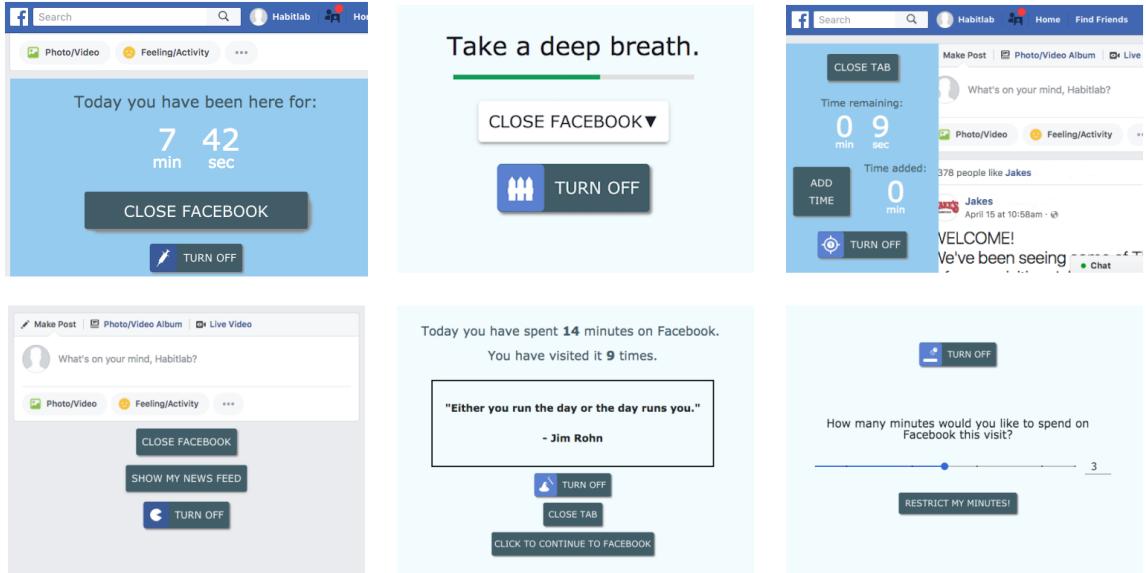


Figure 3.7: Examples of interventions available for reducing time on Facebook. From left to right, top to bottom: a timer injected into the news feed; a page before opening Facebook requiring that the user wait a few seconds before visiting; a countdown timer that automatically closes the tab after time elapses; an opt-in required to show the news feed; an interstitial page before opening Facebook with a quote; an interstitial page before opening Facebook that requires the user set a time limit for how long they will spend this session.

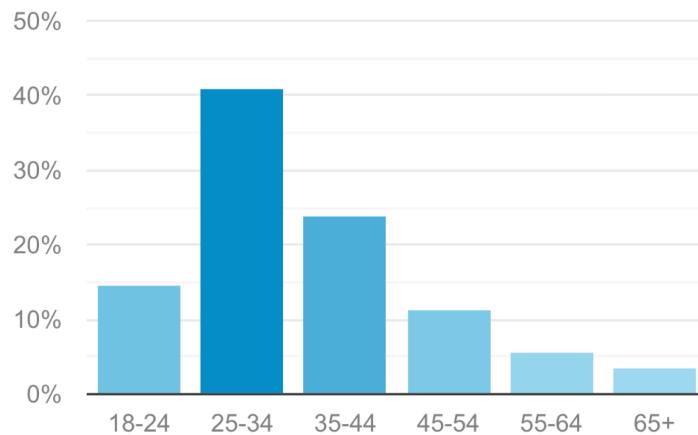


Figure 3.8: Ages of HabitLab users. 25-35 is the most-represented demographic.

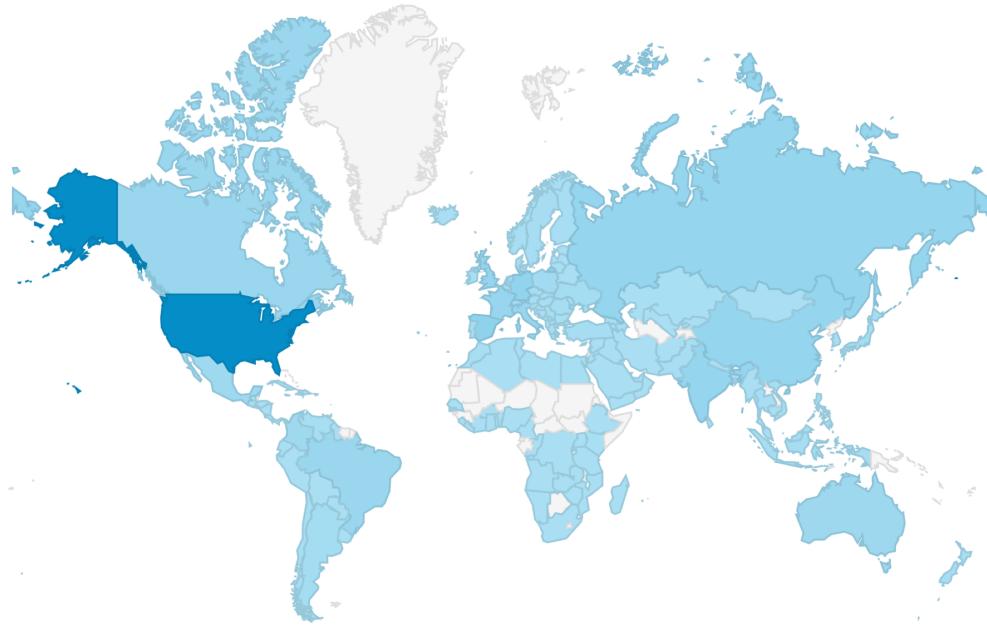


Figure 3.9: Map of countries representing HabitLab users. North America, Europe, and Asia are all well-represented.

German, Greek, Italian, Polish, Portuguese, Russian, Spanish, and Turkish).

The users were not explicitly recruited, but were rather all organic installs who discovered the extension/app via sources such as the Chrome/Play store, or were referred to it via press coverage in sources such as Wired or the New York Times.

Users are asked to read and provide consent to the research protocol upon installation. They may opt out of data collection if they do not wish to have their data analyzed for research purposes.

### 3.5 Design principles and tradeoffs

We designed HabitLab from the start intending it to be a in-the-wild experimentation platform with a large number of users who would voluntarily and organically install it. As a result, we made a number of design decisions that prioritize growth and retention.

Interventions can all be disabled by the end user, either temporarily for the duration of a session via a “Turn off” button shown on each intervention, or permanently. This is intended to boost retention by preventing uninstalls caused by users disliking a particular intervention. While this complicates some analyses – for instance, we may have fewer samples about the effectiveness of less popular interventions that tend to be disabled more – we believe this to be the appropriate tradeoff.

Interventions are designed to be minimally intrusive. While in principle users can just disable interventions they do not like, we still found that many users would uninstall after seeing particular, intrusive

1.	United States	3,710	30.16%
2.	Spain	605	4.92%
3.	Germany	540	4.39%
4.	Russia	461	3.75%
5.	China	450	3.66%
6.	India	448	3.64%
7.	United Kingdom	439	3.57%
8.	France	418	3.40%
9.	Italy	402	3.27%
10.	Canada	368	2.99%

Figure 3.10: Top 10 countries using HabitLab. Users from the US account for 30% of our userbase.

1.	English (US)	6,143	50.00%
2.	Spanish	801	6.52%
3.	English (UK)	725	5.90%
4.	Chinese (Simplified)	462	3.76%
5.	Russian	443	3.61%
6.	Italian	340	2.77%
7.	Portuguese (Brazil)	293	2.38%
8.	French	289	2.35%
9.	German (Germany)	260	2.12%
10.	German	212	1.73%

Figure 3.11: Languages that HabitLab users set as their preferred language to show webpages in. English is the preferred language of half of our users.

interventions. We saw this pattern most notably with interventions that have interstitial screens – that is, they prevent the user from interacting with the page until they have gone through the intervention. As a result, with the exception of a handful of interventions which must be in the interstitial format – for instance, forcing the user to wait for 10 seconds before loading the page – we tried to avoid interstitial interventions as much as possible.

Interventions are designed to load fast, and as a result we ensure that all interventions can work offline and do not depend on remote network resources that might take a long time to load. We found that for interventions that take a second to load or more, the uninstall rate would increase after seeing them. This effect is particularly evident with interstitial interventions, which would have the jarring effect of allowing the user to use the site for a few seconds and disrupting them with an interstitial page once the intervention loads.

We have a simple and short onboarding process. Notably, we do not have long demographic surveys that characterize other similar research projects like LabInTheWild, relying on data from Google Analytics to gather demographic data instead. While data from Google Analytics is only approximate – it is estimated from browsing patterns rather than from asking users directly – we believed that if the data is accurate enough to be used by market research companies worldwide, it would be adequate for our purposes. Furthermore, requiring users to complete onboarding demographic surveys would not be able to guarantee that users would answer truthfully.

This principle of minimizing the amount of questions the user must answer also extends beyond the onboarding process. We make only minimal usage of experience sampling. We do so because we saw in one of our studies that even minimally intrusive, single-click experience sampling prompts that users can safely ignore will significantly increase the uninstall rate. As a result, most of the data we are able to gather is quantitative in nature, and we are only able to gather limited qualitative data from what users report to us through email, our feedback pages, reviews left on app store pages, or the uninstall survey.

### 3.6 List of Browser Interventions

The following is the list of interventions included in HabitLab, showing the intervention name and description as seen by the end user.

Generic interventions that can be used on all sites:

- Minute Watch: Notifies you of time spent every minute
- Supervisor: Shows time spent on site at the top of screen
- Scroll Freezer: Freezes scrolling after a certain amount of scrolls
- Stat Whiz: Show time spent and visit count each visit
- GateKeeper: Makes you wait a few seconds before visiting
- 1Min Assassin: Closes tab after 60 seconds
- Bouncer: Asks how long you want to spend on site this visit

Facebook-specific interventions:

- Time Injector: Injects timer into the Facebook feed
- Feed Eater: Removes the Facebook news feed
- TimeKeeper: Notifies you of time spent in the corner of your desktop
- No Comment: Removes Facebook comments
- Clickbait Mosaic: Removes clickbait from the news feed

Youtube-specific interventions:

- Sidekicker: Remove sidebar links
- Think Twice: Prompt the user before watching a video
- No Comment: Removes comment section

Netflix-specific interventions:

- Fun Facts: Gives you a fact and links an article on the effect of TV
- Alarm Clock: Asks the user to set an alarm before watching a show
- Stop Autoplay: Stops the site from automatically playing the next video

Reddit-specific interventions:

- Comment Remover: Removes Reddit comments
- Mission Objective: Asks what you aim to do this visit and puts a reminder up

Youku-specific interventions

- Think Twice: Prompt the user before watching a video
- Sidekicker: Remove sidebar links

iQiyi-specific interventions

- Think Twice: Prompt the user before watching a video
- Sidekicker: Remove sidebar links

Twitter-specific interventions:

- Feed Eater: Removes the Twitter news feed

Amazon-specific interventions:

- No Recs: Hides recommendations

Gmail-specific interventions

- Speedbump: Delays the arrival of new emails

### 3.7 List of Mobile Interventions

All mobile interventions are generic, that is they can be used on any app.

- At it Again: Sends a pop up with your app visit count.

- Progress Report: Sends a pop up with today's total usage for a certain app
- Red Alert!: Sends a notification with today's total usage for a certain app
- Repeat Offender: Sends a notification with your app visit count
- All in All: Pops a dialog with the day's total time on the current app
- Back To Target: Suggests you to visit a target app
- Counting on You: Puts a timer on screen in watchlisted apps
- Man Overboard! Shows a dialog with your app visit count
- No Peeking!: Asks for confirmation before opening watchlisted apps
- Wait Up! Pause for 10 seconds before entering an app
- Your Better Half: Sends a pop up to go to a target app
- Look on the Bright Side: Dim the screen a little at a time
- Take Your Pick: Select how long you want to spend on an app
- The Final Countdown: On screen timer that closes the app when time runs out

The following interventions apply across the device as a whole, not individual applications.

- How Time Flies!: Sends a pop up message with current app visit length
- Knock Knock: Sends a pop up with your glance count for the day
- Long Time No See: Sends pop up with your phone usage for the day
- Call it a Day: Sends notification with phone usage for the day
- Easy on the Eyes: Sends notification with glance count for the day
- Hello, Old Friend: Sends notification with unlock count for the day
- The Clock is Ticking: Sends a notification with the current app visit duration
- En Garde: Pops a dialog with the day's total unlock count
- Hold the Phone: Show dialog with phone usage for the day
- Long Story Short: Pops a dialog with the visit time for the current app
- Quote reminder: Show quote upon opening app
- Time Reminder: Show dialog with phone usage for the day
- Take Your Pick: Select how long you want to spend on an app

### 3.8 User Feedback

User feedback for HabitLab has been generally positive. On the Chrome store for the browser version, there are 26 reviews, with an average rating of 4.5 stars, while on the Play store for the mobile version, there are 24 reviews, with an average rating of 4 stars. Users leave us feedback, both positive and negative, in a number of forms – through feedback forms within the interface, as shown in Figure 3.12, by filing issues on GitHub, or sending emails. A complete list of feedback that users agreed to have publicly shared is in Appendix E.

We find that most user feedback falls into the following categories:

- Requests:
  - Requests for additional interventions
  - Requests for additional features and ways to customize the system

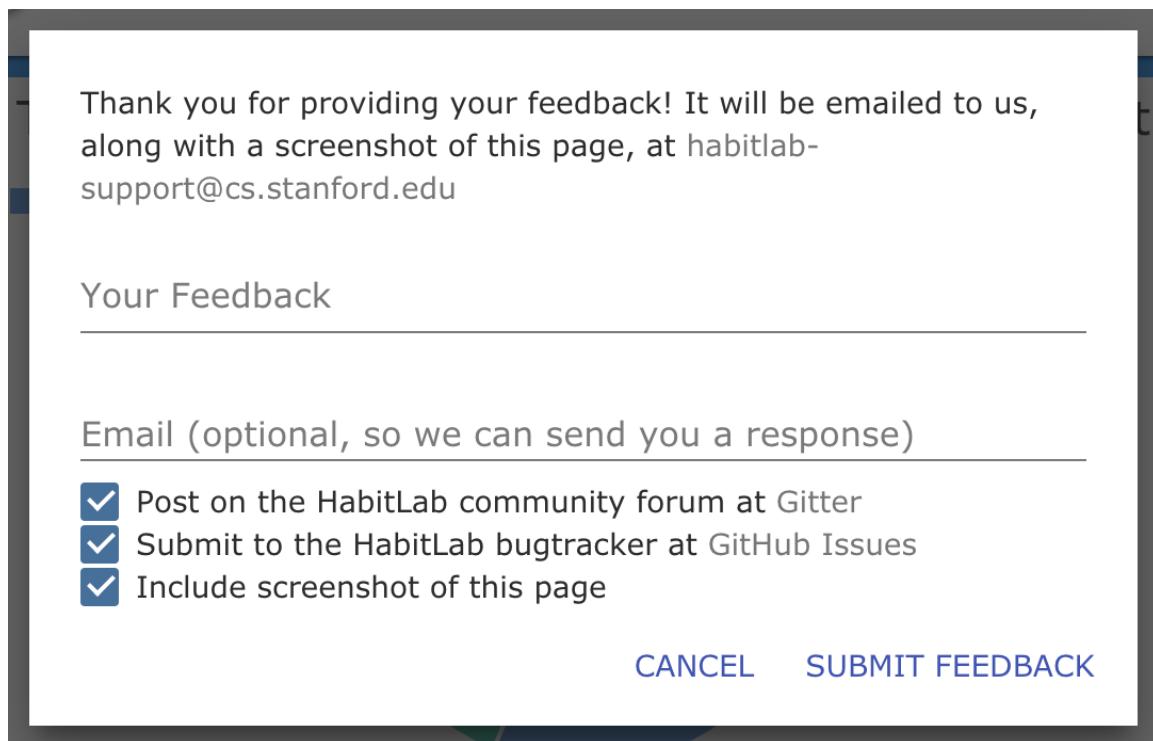


Figure 3.12: Our interface for submitting feedback within HabitLab.

- Requests for additional visualizations in the dashboard
- Requests for site-specific functionality
- Complaints:
  - Complaints about resource usage
  - Complaints about particular interventions
  - Complaints about experience sampling
  - Complaints resulting from A/B testing and experimentation
  - Complaints due to user misunderstanding caused by excessive configurability
- Positive feedback

### 3.8.1 Requests

#### Requests for additional interventions

We find that user feedback often request a specific intervention. Many are site-specific interventions. A commonly requested feature is to have combinations of interventions, or the ability to have multiple interventions active at once. We decided not to go down this path for two reasons:

- There is an exponential number of possible intervention combinations – specifically, there are  $2^n$  subsets of a group of n interventions. Thus, quantifying the effectiveness of combinations of interventions would take considerably more data than quantifying the effectiveness of individual interventions.
- Some combinations of interventions would not work, provide a poor user experience, or would make little sense. For instance, it would make no sense to combine an intervention that injects items into feeds, with an intervention that removes feeds. To avoid deploying these to users, we would have to test and specify which particular subsets of interventions make sense to have together – which, given that there are an exponential number of possible intervention combinations, would take considerable effort.

Some examples of feedback of this form:

*I think it would be great if there was the option for greater control to select multiple nudges to function every time. In my scenario, the website I want to control is Youtube. I use Youtube a lot whilst I'm working and being productive for various tutorials, downloading copyright free assets and so on. However, the recommended videos often push click bait "trending" content at me, and having gone on Youtube to find a tutorial on solving a certain problem, you suddenly find your self 5 minutes into a "You won't BELIEVE what Gordon Ramsey says to this Chef" or similar rubbish. I'd really like to be able to use the Feed Diet, Sidekicker, Supervisor, and No Comment at the same time. I feel like your app has everything I need, but I can't use it all at the same time :)*

– <https://github.com/habitlab/habitlab/issues/641>

*Youtube nudges not working as expected. Sidebar and comments turned on and showing normally. How many nudges can I active simultaneously? –* [\*https://github.com/habitlab/habitlab/issues/620\*](https://github.com/habitlab/habitlab/issues/620)

Users also often request new ideas for interventions. The following are some intervention ideas which were submitted via GitHub. The majority of ideas, however, are submitted via the idea voting interface in

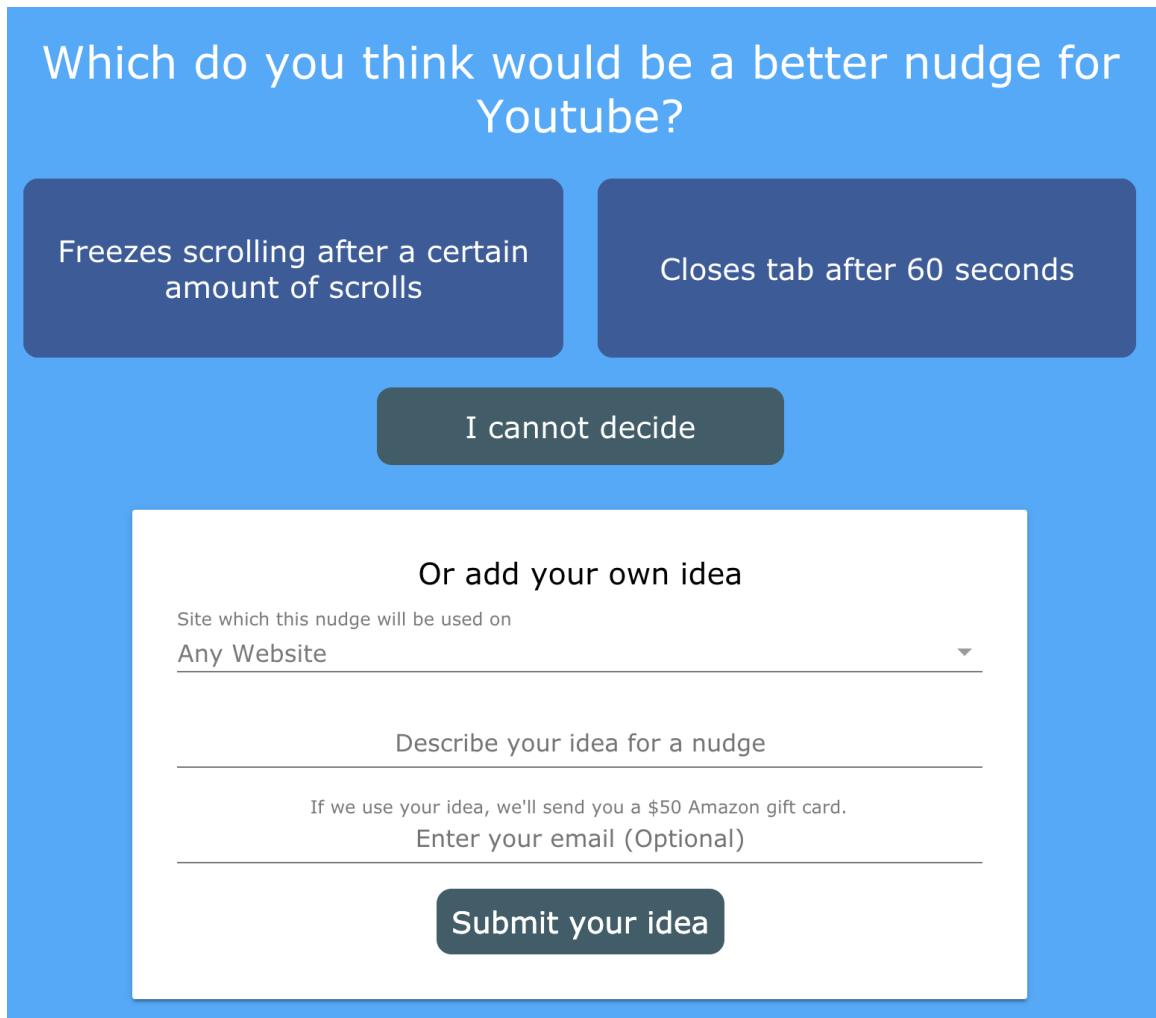


Figure 3.13: Our interface for letting users submit new intervention ideas and vote for existing ones.

HabitLab, shown in Figure 3.13 and are shown in Appendix D.

*i would like to suggest maybe you can make another tracking nudge that only allows you to stay on a certain website and cant open any other tabs – <https://github.com/habitlab/habitlab/issues/593>*

*Hey! Would be nice to have displayed the total amount of time spent on the web. Great work, people! Thank you for making this :) – <https://github.com/habitlab/habitlab/issues/627>*

*Start making font (and other content) fade to grey with each scroll. Slowly the font will become tougher and tougher to read. – <https://github.com/habitlab/habitlab/issues/512>*

*nudge “1 min assassin” that decreases to e.g. 30sec assassin if you already spent 10 minutes on the domain – <https://github.com/habitlab/habitlab/issues/536>*

### Requests for additional features and ways to customize the system

Another commonly-seen request is for additional features and the ability to customize certain aspects of the program.

A commonly requested feature is the ability to turn off the “Turn Off” button which is present in each intervention.

*Option to hide “Turn off HabitLab” button. <https://github.com/habitlab/habitlab/issues/605>*

*Please add an option in settings to hide nudge turn off buttons (so they can only be turned off from settings page). – <https://github.com/habitlab/habitlab/issues/604>*

*Please allow option to not turn off buttons, e.g. “turn off feed eater button”. Ideally this is a global option across all extensions. Thanks! :) – <https://github.com/habitlab/habitlab/issues/361>*

Some users request more flexibility in which interventions are shown in particular contexts:

*In general, I think the random is a good idea, but I quickly realize some of the random ones are ineffective for me. Rather than the random, it would be great if I could designate 1 feature for a particular website. For example, I know that facebook is a big time waster to me and moreso than others. I would love it if that was 1 one minute kick-off no matter what. Others (eg - Twitter) are less of a distraction for me, so the random wouldn’t be as painful. – <https://github.com/habitlab/habitlab/issues/583>*

*have multiple slots for work times. eg : 8h00 12h00 and 14h00 18h00 – <https://github.com/habitlab/habitlab/issues/533>*

Some users request more flexibility in specifying sites to reduce time on, beyond our blacklist of domains model. One model frequently requested was a whitelist model:

*I really wish there was a setting where I could use a nudge on every website except what’s on a whitelist, because I always find a new place to waste time. – <https://github.com/habitlab/habitlab/issues/637>*

*Please add a whitelist for stricter management, so we add the sites we DO want to access. – <https://github.com/habitlab/habitlab/issues/548>*

*Can you make a whitelist version? <https://github.com/habitlab/habitlab/issues/553>*

Another model users suggested would be to group and categorize sites, and limit time across site categories:

*You should be able to group sites and provide overall limits and timers across the category. For instance, Netflix and YouTube would be considered ‘media streaming’ allowing the user to set a goal of say an hour of streaming a day. Otherwise, I can personally see me spreading my viewing over multiple different websites to (pretend to) keep within the goals. – <https://github.com/habitlab/habitlab/issues/632>*

Some users want more fine-grained ways to specify where to spend less time, beyond just the level of domains:

*The extension doesn’t track subdomains of a main domain properly. Suppose I add xyz.com as a filter and then it directs me to abc.xyz.com, the filter doesn’t provide the needed nudges. This can be helpful if implemented. – <https://github.com/habitlab/habitlab/issues/566>*

*What about separating out Amazon from its Kindle page ([read.amazon.com](http://read.amazon.com)) and its music page ([music.amazon.com](http://music.amazon.com)) because I enjoy listening to music when I type, and (sometimes) I read books on my Kindle from the website for work.* – <https://github.com/habitlab/habitlab/issues/633>

*On sites like [news.google.com](http://news.google.com) and [reddit.com](http://reddit.com), you can click on links that take you to long articles on other domains where you can spend a significant chunk of time. Habit Lab doesn't track such changes now but really should!* – <https://github.com/habitlab/habitlab/issues/601>

*I want to be able to temporarily turn off nudges for a particular website. For example, I'm watching educational youtube videos, but still want to avoid other sites.* – <https://github.com/habitlab/habitlab/issues/522>

### **Request for additional visualizations in the dashboard**

Many users request additional information to be shown in the dashboard which we have the data for, but do not show any visualization for:

*I would love it if it would give me a running tally of my total minutes spent in addition to the "Today's five most visited sites by minutes spent" Thanks and cheers!* – <https://github.com/habitlab/habitlab/issues/562>

*I want to see my history and detailed results for longer durations, such as weeks or months* – <https://github.com/habitlab/habitlab/issues/618>

*I want to see results per week and month - and to be able to compare each week / month. Also, an option to choose which day the week starts* – <https://github.com/habitlab/habitlab/issues/597>

*How much time did I spend on what tabs, last week? Where are the totals?* – <https://github.com/habitlab/habitlab/issues/561>

*Show total browsing time including those pages not in top 5.* – <https://github.com/habitlab/habitlab/issues/607>

### **Requests for site-specific functionality**

Other requests for features and customizations appear to be some specific to individual sites. Some examples are shown below:

*Please let me have the "Mission Objective" nudge on youtube as well! I feel like it's one of the most effective nudges and would really make me consider twice whether to watch youtube or not.* – <https://github.com/habitlab/habitlab/issues/544>

*The disabling of autoplay makes me skip the recap of Jane The Virgin. However, that series includes new information and new jokes in every recap, so I always watch each recap even if I've just seen the previous episode. With autoplay on I can't watch the recap even if I've just logged in to Netflix, and even if I'm "rewinding" back to the beginning of the episode each time. It skips it again.* – <https://github.com/habitlab/habitlab/issues/639>

### 3.8.2 Complaints

#### Complaints about resource usage

HabitLab is a complicated codebase with many background tasks, so it results in additional resource usage which may cause some users who monitor browser resources to uninstall it:

*Disabling HabitLab due to excessive CPU usage. Right now the Chrome browser's TaskManager, CPU usage column shows HabitLab using 10-15% CPU. Ouch! Adios! – <https://github.com/habitlab/habitlab/issues/634>*

*HabitLab extension in Chrome is using a lot of CPU – <https://github.com/habitlab/habitlab/issues/565>*

#### Complaints about particular interventions

We have several complaints directed towards particular interventions. The more intrusive interventions, which may prevent the user from using some functionality on the site, in particular have more complaints directed towards them. Some examples of feedback of this form follow:

*I came to Facebook to check notifications for events, but the scroll freezer hides the entire top bar. And the search bar too! I can't manage the events that I came here to manage. The other HabitLab stuff is good though! – <https://github.com/habitlab/habitlab/issues/617>*

*Nudges should NOT cover the page, they should possibly push the whole page down by the amount of space needed by the nudge. Most apps, like twitter and facebook, have their buttons at the top, and your nudges simply cover those and make the site unusable, so maybe one wants to do a quick action on the sites bar and leave, but the nudge bar is in the way, so one is forced to close the nudge to access the function of the site and get on with it. But then the nudge is closed – <https://github.com/habitlab/habitlab/issues/613>*

*Banner is too large – <https://github.com/habitlab/habitlab/issues/571>*

*I turned the restrict your time nudge off because it caps at 5 min. I like the idea, but it needs to be free form. – <https://github.com/habitlab/habitlab/issues/446>*

#### Complaints about experience sampling

Experience sampling was a HabitLab feature that led to a number of complaints, despite our efforts to ensure they were as unintrusive as possible. Examples are shown below:

*Whenever I go to a “nudged site”, this “how aggressive” overlay comes up. I would not like it to. I click on “Light touch” every time, and it’s so annoying that I’d sooner remove the Chrome extension than keep doing it every time. I get the thought, that maybe the annoyingness will make me visit those sites less. But if I wanted to not visit them at all, I’d just block them outright using another extension. I want to visit them, but be aware of how much time I’m spending. And I don’t want additional tasks to accomplish every time. Would the “how aggressive” panel triggering-or-not be a setting you could add? I’d really like to keep using this system! – (via email)*

*How do I disable the “How aggressive would you like HabitLab to be in helping you reduce your time spent this visit?” message when going to Facebook? I find myself mindlessly clicking “don’t do anything”.... would prefer to have nudges on by default without an option to determine the strength BEFORE each FB visit ... this seems to be a new feature, that enables me to spend more time on FB without nudges ... how do I disable this??? TIA. – [https://www.reddit.com/r/habitlab/comments/a9kvly/how\\_do\\_i\\_disable\\_the\\_how\\_aggressive\\_would\\_you/](https://www.reddit.com/r/habitlab/comments/a9kvly/how_do_i_disable_the_how_aggressive_would_you/)*

*it keeps asking me “how much do you want me to bother you??” and i am tired of answering this question. very cool extension that i used for like a year but something seems to have gone wrong so now i’m uninstalling :( – <https://github.com/habitlab/habitlab/issues/638>*

*Feed Eater Bug- if I have the feed eater feature enabled, every time I open facebook, there will be an alert window along the lines of “How aggressive would you like HabitLab to be in helping you reduce your time spent this visit?”, which gets tiring when you have to open facebook a lot of times for personal matters (not time wasting stuff I swear). Update: Actually ignore what I just said about the Feed Eater bug, even with the feature off the issue continues on. – <https://github.com/habitlab/habitlab/issues/547>*

### **Complaints resulting from A/B tests and experimentation**

Some complaints were due to users misunderstanding how certain features of HabitLab work, which they perceived as being bugs.

One feature which frequently led to confusion was the rotating nature of interventions. Many users were expecting to see the same intervention every visit:

*This is one of the most useful extensions when it comes to fighting web addiction. However, many of the nudges, such as the news eater for Facebook fail to work at some occasions. And most of them does not work when combined with other nudges. This often defeats the purpose of the extension entirely. Otherwise, I love the idea of having nudges, the pie chart for an overview and setting daily limits. In the meantime however, I will use News Feed Eradicator for Facebook, WasteNoTime and RescueTime instead. – <https://github.com/habitlab/habitlab/issues/511>*

*Sidekicker, and NoComments are not working on YouTube.com. They work on “Try now” mode, but when actually expecting it to run while browsing, it does not work. I can see see Comments and Side bar. – <https://github.com/habitlab/habitlab/issues/538>*

*Hello! I am not sure if I am understanding properly, but with the Bouncer nudge (my favourite), once you have done it once in a day it never triggers again. It would be good for it to ask every time I go to a site, how long I want to spend on it. And if I exit the site, it refreshes and starts again. – <https://github.com/habitlab/habitlab/issues/576>*

Sometimes intentional artifacts of our A/B tests led users to believe that the system was broken or buggy. For example, in one of our A/B tests we varied the frequency at which interventions were shown, so often users would not see an intervention. This resulted in many users reporting bugs that they were not seeing interventions, even though this was intentional:

*Why do I not always receive a nudge when I visit facebook, I keep compulsively checking, and hoping a nudge will remind me, but I don't seem to be seeing any. – <https://github.com/habitlab/habitlab/issues/557>*

*Nudges will frequently not show up when I visit facebook. Why is this? I haven't accidentally turned them off. – <https://github.com/habitlab/habitlab/issues/572>*

*the nudges aren't really working....i have visited facebook multiple times now, but not have been nudged even once. I have enabled all the nudges as to see which one helps me the best, but its not nudging at all. – <https://github.com/habitlab/habitlab/issues/570>*

*None of the nudges are showing up. I'm not sure how to track down the issue. I've pasted the contents of the javascript console from a visit to YouTube here <https://pastebin.com/ZyPYB2Tw> in case it helps track down the issue. Does HabitLab not work alongside adblockers or ghostery or something like that? – <https://github.com/habitlab/habitlab/issues/569>*

*My nudges are not working most of time. – <https://github.com/habitlab/habitlab/issues/594>*

*It won't show nudges – <https://github.com/habitlab/habitlab/issues/585>*

### **Complaints due to user misunderstanding caused by excessive configurability**

Certain aspects of our system design caused user confusion and complaints. The most prominent ones were intervention rotation, and artifacts of A/B testing, which we described in separate sections above. Some others that resulted from a misunderstanding of functionality and controls are shown below:

Interventions are per-site, so if a user turns off an intervention on one site, it will not automatically turn off that intervention on other sites. This caused some users to believe that their request to turn off interventions was being ignored, when in reality they just had to turn it off on all sites. This suggests perhaps we should prompt the user whether they wish to turn off an intervention on all sites, when they disable it on one site, or simplify the configuration to default to turning interventions off on all sites by default:

*I have told this app again and again that I want to turn off 2 nudges, freeze scrolling, and setting the number of minutes I want to spend on a website in advance. HOWEVER IT KEEPS COMING BACK and I am almost about to delete this despite loving ever other aspect of it. FIX THAT ASAP. OR don't say you can remove it if you can't. – <https://github.com/habitlab/habitlab/issues/628>*

To satisfy various requests we received during development to be able to temporarily turn off interventions, we had options to turn off interventions for just that particular visit, for the entire day, or permanently. This sometimes led to confusion, as users would select the option to turn off an intervention temporarily, but they actually had intended to turn off the intervention permanently:

*turned on for twitter - minute watch distracting / annoying, so turned on (?) "supervisor", turned off minute watch. It doesn't take. It still annoys. I'm about to simply turn off habit lab as a result. – <https://github.com/habitlab/habitlab/issues/522>*

To address such confusion, We added some dialogs when turning off interventions to help explain what

they would do, and provide alternative options in case they were looking to turn off interventions permanently or on all sites – though this itself resulted in complaints:

*When I turn off for the day/for the rest of the visit, it would be nice not to have a modal to confirm : I need one more click to close it, and it's annoying. If I wan't to turn it off, then there is no use anymore to slow down my use of thoses websites. See <https://modalzmodalzmodalz.com/> for help – <https://github.com/habitlab/habitlab/issues/517>*

Thus, user preferences are also extremely varied. In particular, users constantly request additional ways to customize and configure the system. However, if we add the ability to customize the system at too high a granularity, this may increase user confusion, as we observed with users intending to turn off an intervention permanently on all sites being frustrated that they were still seeing that intervention, once we added options to turn them off per-site or just temporarily that they mistook for turning them off permanently on all sites.

### 3.8.3 Positive Feedback

Some users also left us positive feedback:

*Guys, I fucking love what you made! I send you all my gratitude for your wonderful product! – <https://github.com/habitlab/habitlab/issues/404>*

*YOU GUYS ARE AWESOME! I love nudges. I love Stanford. This helps me so much. I wish everybody knew about it. – <https://github.com/habitlab/habitlab/issues/406>*

## 3.9 Uninstallation Survey Responses

The HabitLab system also includes an uninstallation survey that users can optionally answer when they uninstall. We show a screenshot in Figure 3.14.

We asked users their reasons for uninstalling. Users could select more than one of our predetermined categories, and leave free-form feedback if their reason did not fit into any of our categories. Of a total of 4635 users who responded to our uninstallation survey, they chose 5782 reasons. A breakdown of uninstallation reasons is shown in Figure 3.15. The most common reasons for uninstalling concerned the interventions themselves – interventions being too annoying was the most commonly selected reason, followed by interventions being considered as ineffective. A complete list of free-form feedback users left about their reasons for uninstalling is shown in Appendix F.

## 3.10 Discussion

The HabitLab system was built on the premise that experimentation for behavior change can be useful for both end users as well as researchers. Based on our experiences building, growing, and receiving feedback on the system, we have reasons to believe that this premise is true in certain respects, while false in others.

## Sorry to see you go! HabitLab has been uninstalled

Thanks for trying HabitLab! We'd appreciate your feedback so we can make it better! Why are you uninstalling?

- Interventions were annoying**
- Was causing lag**
- Did not feel effective**
- Privacy concerns**
- Other reason**

We would appreciate your feedback here

Submit

Figure 3.14: The survey shown to users if they uninstalled.

Reasons for uninstalling, among users who submitted the uninstallation survey

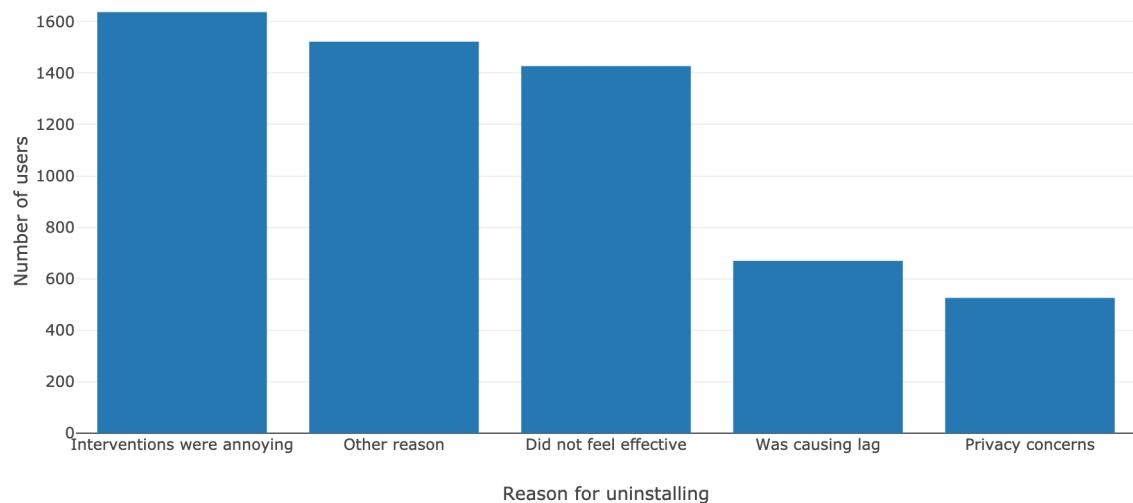


Figure 3.15: Reasons users selected for uninstalling, in responses to our uninstallation survey.

Specifically, certain types of experimentation can be well-aligned with end user goals and perhaps even enhance the user experience, but other types of experimentation may be detrimental to the user experience.

We observed that among negative feedback we received for the system, the most common complaint related to experience sampling – users simply did not like being forced to answer questions, regardless of how minimally intrusive they were. The next most common complaints were necessary artifacts of our experimentation and A/B testing – some users wanted more control over the frequency of interventions and which interventions were seen at particular times, but we wanted to randomize these as part of our experiments. We also often found that users requested additional customizations that would provide them with more control over interventions, but were reluctant to implement them as they would complicate experimentation by making it more difficult to quantify intervention effectiveness.

We also find that users may be unsatisfied with “baseline” or “control” conditions in an in-the-wild context. Specifically, some of our experiments necessitated that users occasionally not be shown any intervention on a particular visit, so that we can observe baseline time spent in the absence of an intervention. However, whenever we had such an experiment running, this resulted in a flurry of user complaints that interventions were not working. Even when users were aware of this, or were aware of the rotating nature of interventions, they expressed dissatisfaction that they were not able to see particular interventions each visit. Thus, this is a clear example of researchers’ needs for experimentation being in conflict with users’ preferences.

That said, several forms of experimentation did not conflict much with user expectations. In contrast to experience sampling, which resulted in both user complaints and increased attrition, we found that adding the same question as a one-time onboarding question did not result in user complaints or increased attrition. Likewise, while we found that the intervention rotation strategy increased attrition and resulted in some user complaints, it also increased intervention effectiveness.

Thus, while we would optimally always be able to find experiments which both satisfy scientific needs as well as preferences of end users, they are often in conflict. In-the-wild experimentation platforms such as HabitLab must sometimes make compromises between these goals. With HabitLab, our approach to navigating this problem has tended to be experimental. Specifically, we A/B test our experiments and constantly monitor user feedback and attrition rates in experimental conditions compared to baselines which we know to have good user satisfaction. If a particular experiment is causing a large increase in attrition, or users are unsatisfied, we attempt to modify the experiment design so we can answer the same research questions with less increase in attrition. For instance, since we found that our original experience sampling prompt design causes an increase in attrition, we A/B tested several different variations to try to find alternative designs which would collect similar data while causing less attrition.

Our general approach to this experimentation-vs-user-experience tradeoff with HabitLab can thus be described as using A/B testing so we can understand the tradeoff we are making, and looking for alternative experiment designs that satisfy our scientific goals with better user experience if they are in conflict. We believe that designers of in-the-wild experiment platforms must always be thinking about user experience,

retention, and satisfaction, and cannot sacrifice these for the sake of running an experiment – after all, an in-the-wild experiment platform’s most precious asset is its userbase, and if there are no users, we cannot run any experiments.

## **Chapter 4**

# **Rotating Online Behavior Change Interventions**

### **4.1 Introduction**

Excessive recreational online activity, or “cyberslacking” is a major problem. Hence, numerous productivity interventions, such as time trackers and site blockers, have emerged to help users reduce their time online. Existing productivity tools typically always show the same intervention; however, techniques that alternate between different interventions (selected via algorithms such as multi-armed bandits) have been found to improve overall effectiveness in various health intervention contexts. This can be applied to online productivity tools as well – i.e., hiding the news feed on one visit, injecting a timer into the page on others, etc.

Alternating interventions has typically been done in the context of an adaptive intervention selection algorithm (e.g. multi-armed bandits), and the benefits have been attributed to personalization – certain interventions are more effective for certain individuals, and by exploring which interventions are most effective for the individual, the algorithm can maximize effectiveness in the long run.

However, we believe that our understanding of the effects of alternating interventions may be incomplete. Certain health interventions have been found to decline in effectiveness over time, and banner blindness is a well-known phenomenon in online advertising – if users analogously develop blindness to interventions, could the novelty brought by alternating interventions itself improve effectiveness, by preventing the user from developing intervention blindness? Such effects are important, because declines in intervention effectiveness over time violates a major assumption of multi-armed bandit algorithms – that the rewards, e.g. the effectiveness of the interventions – do not change over time. If alternating interventions has a significant positive effect on effectiveness, intervention selection algorithms need to take this into account.

Alternatively, alternating interventions can have negative effects – users may be annoyed and fatigued if they constantly see different interventions, as it may violate expectations of a consistent user experience. Users may also feel a lack of control if the system is choosing which interventions to show them, rather than always showing the same one. If a user particularly dislikes one of the interventions, the user may be exposed to it while the system is exploring the space of interventions, which risks causing user dissatisfaction. These factors may lead to alternating interventions contributing to increased rates of attrition – which intervention selection algorithms need to take into account.

To answer these questions of in which ways alternating interventions can be beneficial or harmful, we developed a Chrome extension, HabitLab, that features a number of online productivity interventions to help users reduce their time spent on sites like Facebook. We released it publicly on the Chrome web store, and ran a pair of in-the-wild studies on users who newly installed the extension, where we varied how much we were alternating between interventions. Users were unpaid, organic installs from the Chrome web store rather than artificially recruited participants, which allows us to better observe real-world usage and attrition patterns. We found the following results:

1. Users who are constantly rotating between interventions have higher rates of attrition than users who consistently see the same intervention.
2. Interventions decline in effectiveness over time
3. Interventions are more effective at reducing users' time spent when they are shown alternating between different interventions
4. Users are less likely to attrite if they chose their interventions themselves during the onboarding process

These results led us to hypothesize that the increase in attrition observed due to alternating interventions may be a result of violation of user expectations (due to users having an incorrect mental model), or users feeling a lack of control.

To investigate the underlying cause and mitigate this attrition, we developed a pair of interface techniques shown when a new intervention is seen for the first time. The first interface just repeats information (that users saw during onboarding but may not have read) that the system may show a different intervention on each visit, while the second interface gives users a sense of control by allowing them to opt-out of seeing the intervention in the future. We found that adding these interventions both significantly decrease attrition.

Behavior change systems today suffer from declining effectiveness as novelty wears off over time. Typically, behavior change systems utilize a *static* design, which never changes. For example, to manage social media browsing time, the three popular options are to block tempting sites [160], use a work timer [55], and audit time spent [83, 132]. However, static interventions suffer from high attrition and abandonment rates [32, 51], and interventions decline in effectiveness over time [91]. Habituation eventually drives users

to stop paying attention to, or avoid, static interventions—an effect often seen on the web as banner blindness [20]. The end result is that many behavior change systems are tuned out by users, and are unsuccessful at their goals.

If static interventions are tuned out, *rotation* might provide a remedy. Much like a human coach or tutor rotates between different approaches over time, rotation might maintain attention in ways that static interventions cannot. Online behavior change tools could apply similar techniques, for example injecting a visible stopwatch timer into the page on one visit to Facebook, and hiding comments on the next visit. Techniques that personalize interventions via multi-armed bandits show positive treatment effects, suggesting that the approach may hold promise, but this existing work cannot separate the effects of personalization from the effects of rotation [129, 92, 119]. Because rotated interventions continually change the user interface, however, they may frustrate users by violating consistency and a sense of user control, leading to lower effectiveness or higher attrition.

This chapter takes up the question: are static or rotated interventions more effective for behavior change? Is it possible to understand the effects of rotation in order to design more effective behavior change systems? We focus specifically on helping users who want to manage their time on social media websites such as Facebook, YouTube, Reddit, and Twitter. We perform a series of field experiments with people who sought out and installed a browser extension that we developed for online behavior change.

Our platform, *HabitLab* (<https://habitlab.stanford.edu>) is a Chrome extension that features a number of online productivity interventions to help users reduce their time spent on sites such as Facebook. We released HabitLab publicly on the Chrome web store, where, as of the time when we ran these studies, it had attracted over 8,000 daily active users. This user base allows us to observe real-world usage and attrition patterns over time.

We ran three in-the-wild studies on users who newly installed HabitLab. In Study 1, we compared static interventions to rotation. We measured effectiveness through time on the user’s targeted site, and we measured attrition by tracking when users stopped using the extension. Results indicate that rotation is a double-edged sword. Rotating interventions reduced time spent on sites by 34% per day, but at the cost of nearly doubling attrition levels.

Study 2 replicates the first experiment over a longer period of seventy days, and additionally tests whether the number of interventions included in the rotation impacts attrition. The results successfully replicated the original results over a longer 70-day period, and suggested that the larger the set of interventions, the higher the probability of attrition.

To investigate the underlying causes of attrition and mitigate the effects of rotation on attrition, we analyzed user feedback and developed a pair of interface techniques to improve the user experience in the presence of rotation, which we deployed in Study 3. The first technique is informational, aiding people’s mental models by reminding them that the system may show a different intervention on each visit. The second technique focuses on user control, providing the same information as well as a just-in-time mechanism for people to opt out of each new intervention as they see it. Results indicated that these interventions

reduced attrition by over half, so that 80% of new users were still using the system actively after a week.

In sum, this chapter contributes the first comparison of static and rotation intervention strategies in behavior change, a living laboratory system that allows us to deploy this investigation and other field experiments, and interaction design strategies that can help offset increased attrition due to rotation. Its results suggest that people may be more able to control their social media usage more effectively than using today's common techniques such as site blockers. The rest of the chapter is organized as follows: we first review studies of behavior change to develop our research question and hypotheses; we then describe our studies and results; we close with reflections and future design directions.

## 4.2 Related Work

In this section, we review literature in behavior change systems and psychology to develop specific testable predictions regarding the research question.

### 4.2.1 Effectiveness over time

While behavior change systems can be effective [16, 39, 164], many review papers are more restrained in whether behavior change systems remain effective over long periods of time [111, 22, 114, 61]. The critique holds that behavior changes are long, complex processes, and the effectiveness of a system is hard to maintain indefinitely [125]. Prior work suggests that the effectiveness of showing a static intervention cannot be maintained indefinitely [65, 134]. For example, when a health behavior change system started sending email reminders, the first reminder was successful 28% of the time, but by the fifth reminder it was successful only 18% of the time [79]. A further meta-analysis of 88 computer-tailored interventions for health behavior change suggested that the efficacy of interventions decreases over time [91].

### 4.2.2 The impact of rotation

Novelty can be a driving factor for effectiveness. One study showed that novelty can influence encoding of information into long-term memory, which, in turn, may raise awareness of behavioral changes [89]. Studies of gamification also explore the effect of novelty on user engagement [60].

In web design, people begin ignoring parts of the screen that have little information scent, such as ads. This phenomenon is termed banner blindness, after the commonness of the effect in internet banner advertising [20]. As static interventions remain deployed, they may suffer from the same banner blindness and lack of novelty (wear-out) effects, suggesting a potential mechanism for the decreased effectiveness over time.

Rotating interventions may counter these effects. Different interventions appear in different parts of the interface, making it less likely that the user would ignore them wholesale. Online behavior change systems that use machine learning algorithms such as multi-armed bandits hone in on a small number of

interventions to use [119, 127], but during the early exploration phases they are essentially rotating between interventions. Systems that personalize interventions [79] or deploy many micro-studies [43] have generally found positive effects.

#### 4.2.3 Attrition

Attrition is a major challenge in behavior change systems: a metastudy of eHealth interventions found that an attrition rate around 99% over a 12-week period is normal [51]. Likewise, the number of users in a stress-coping mobile application declined in a steady rate through the study [119].

Though rotating interventions aids novelty, the literature suggests that it may hurt attrition. Rotation violates usability heuristics such as consistency and user control [113]. Specifically, users may perceive a loss of control when they are presented with ever-changing interventions, potentially leading to non-compliance behaviors and a higher attrition rate [64]. Typically, in attrition-risky domains such as education, an effective user-centered design is critical for minimizing attrition [14].

### 4.3 Research Questions

A recent trend in behavior change systems has been the concept of personalizing interventions. Such systems explore several possible strategies using techniques such as multi-armed bandits to find the intervention that is most effective for the user [119, 127]. For example, PopTherapy demonstrated personalized messaging could be found through such techniques [119]. Likewise, HeartSteps conducted tens or hundreds of micro-randomized trials on users [43]. When multi-armed bandits are just beginning to get feedback from a user, they will try out several different interventions to see what works. This exploration has the effect of rotation, but the amount of rotation declines as the bandit begins to personalize. In this chapter, we examine the contrarian assertion that perhaps rotation should be maintained to sustain novelty even after the multi-armed bandit is aware of which intervention is most effective for the user.

The challenges of static interventions, and the rising wave of personalization systems, call into focus: would a rotation strategy work? Or is it a weak palliative with little discernible effect? This led to our research question:

**Research Question (RQ).** *Can a strategy of rotating interventions produce more effective behavior change systems?*

While behavior change systems can be effective [16, 39, 164], many review papers are more restrained in whether behavior change systems remain effective over long periods of time [111, 22, 114, 61]. The critique holds that behavior changes are long, complex processes, and the effectiveness of a system is hard to maintain indefinitely [125]. Prior work suggests that the effectiveness of showing a static intervention cannot be maintained indefinitely [65, 134]. For example, when a health behavior change system started

sending email reminders, the first reminder was successful 28% of the time, but by the fifth reminder it was successful only 18% of the time [79].

A further meta-analysis of 88 computer-tailored interventions for health behavior change suggested that the efficacy of interventions decreases over time [91]. This prompts our first hypothesis:

**Hypothesis 1 (H1).** *Static interventions will suffer from decreased effectiveness over time.*

Novelty can be a driving factor for effectiveness. One study showed that novelty can influence encoding of information into long-term memory, which, in turn, may raise awareness of behavioral changes [89]. Studies of gamification also explore the effect of novelty on user engagement [60].

In web design, people begin ignoring parts of the screen that have little information scent, such as ads. This phenomenon is termed banner blindness, after the commonness of the effect in internet banner advertising [20]. As static interventions remain deployed, they may suffer from the same banner blindness and lack of novelty (wear-out) effects, suggesting a potential mechanism for the decreased effectiveness over time.

Rotating interventions may counter these effects. Different interventions appear in different parts of the interface, making it less likely that the user would ignore them wholesale. Online behavior change systems that use machine learning algorithms such as multi-armed bandits hone in on a small number of interventions to use [119, 127], but during the early exploration phases they are essentially rotating between interventions. Systems that personalize interventions [79] or deploy many micro-studies [43] have generally found positive effects.

Based on these results, non-static interventions may be effective. We hypothesize:

**Hypothesis 2 (H2).** *Rotation will increase effectiveness, compared to static interventions.*

Attrition is a major challenge in behavior change systems: a metastudy of eHealth interventions found that an attrition rate around 99% over a 12-week period is normal [51]. Likewise, the number of users in a stress-coping mobile application declined in a steady rate through the study [119].

Though rotating interventions aids novelty, the literature suggests that it may hurt attrition. Rotation violates usability heuristics such as consistency and user control [113]. Specifically, users may perceive a loss of control when they are presented with ever-changing interventions, potentially leading to non-compliance behaviors and a higher attrition rate [64]. Typically, in attrition-risky domains such as education, an effective user-centered design is critical for minimizing attrition [14]. In light of these results, we hypothesize:

**Hypothesis 3 (H3).** *Rotation will increase attrition, compared to static interventions.*

## 4.4 Experiment Platform: HabitLab

We conducted the studies in this chapter using the browser version of HabitLab. At the time the studies presented in this chapter were conducted, there were 8000 daily active users using the platform. The list of

interventions that were included in HabitLab at the time of this study is included at the end of this chapter.

## 4.5 Study 1: Field Study on the Effect of Rotating Interventions

Our first study is a within-subjects design run on the HabitLab platform that aims to understand the effects of rotating interventions on effectiveness and attrition.

### 4.5.1 Participants

Participants in our first study consisted of new HabitLab users installing the system over a period of three weeks in March and April 2018. 692 users installed HabitLab over the course of our experiment and consented to our research protocol. We discarded participants who were not new users of HabitLab, since some users were reinstalls or new devices for existing users. We also discarded participants who did not complete the onboarding process, or who uninstalled the system before they saw their first intervention. This left us with 217 participants.

We do not administer a demographic survey at install time, because long onboarding processes had previously led to high abandonment. Many users find HabitLab through routes other than the web site, but Google Analytics on the web site can provide some window into rough trends. Google Analytics estimates that 89% of visitors to the HabitLab website during the experiment period were male, indicating a male skew. The most common age group was 25–34 (41%), followed by 18–24 (29%), 35–44 (22%), and 45–54 (7%). According to users’ IP addresses, the most highly represented countries were the US (23%), India (12%), Germany (9%), France (5%), and the UK (4%).

Participants agreed to our informed consent protocol during onboarding. This consent protocol indicated that HabitLab would be selecting and rotating between different interventions, but did not mention any specific algorithm or rotation schedule.

### 4.5.2 Method

Participants used HabitLab in the course of their usual web activity. As they browsed, HabitLab would introduce interventions when appropriate. All interventions were available to all conditions, but the pace at which old interventions were replaced by new ones depended on condition. Users would react to the intervention, or not, as they browsed.

HabitLab operated on all web sites that the user had selected upon installation. However, because users spend differing amounts of time on different domains, and there was a long tail of domains which were set as goal sites by only a few users, we restricted analysis to domains where we had a substantial dataset, specifically: Facebook, Youtube, Reddit, Twitter, VK, and Amazon.

The experimental unit was a participant assigned to a condition for a block of days. Block lengths were randomized between 1 day, 3 days, 5 days, and 7 days, in order to give us insight into the effects of rotation

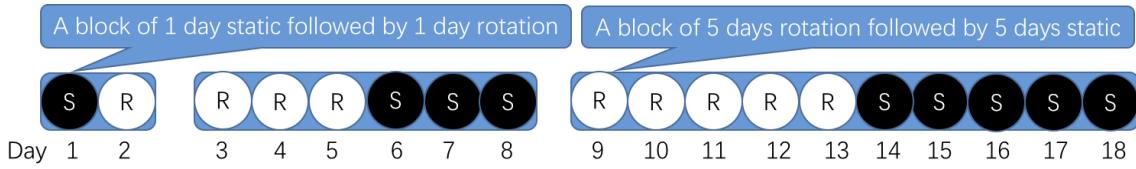


Figure 4.1: Example order in which a user might see conditions. Each circle represents a day – on black days the user is in the “static” condition, white is the “rotation” condition. The order of blocks is randomized; here, this participant is seeing blocks in order 1, then 3, then 5, then 7 (omitted in the figure).

strategy over different time horizons. When participants were randomized to a block, for example a five-day block, they experienced HabitLab in one condition for five days, then in the other condition for five days, for a total of ten days (Figure 4.1). Condition order was randomized within each block. At the conclusion of a block, the user was then moved into another block length and the trial repeated. The sequence of block lengths was randomized for each participant. If they kept the system installed, participants would experience all blocks after 32 days.

#### 4.5.3 Conditions

We developed a within-subjects repeated measures design, where users alternated between blocks of time during which they were shown either a static intervention or rotated interventions. A within-subjects design such as this allows us to better control for the large variability across users in how much time they spend on a site.

The static condition captures a typical behavior change design with one strategy. At install time, for each site, the participant is randomly assigned a single intervention among the ones that are enabled for the site. Whenever they visit that site on a day in the static condition, the participant will always see that intervention, i.e. the static intervention is the same across all blocks.

The rotation condition captures a strategy of keeping the interventions changing so that users do not begin ignoring them. Each time a participant in the rotation condition visits a target site (e.g., Facebook), HabitLab picks a random intervention from the enabled set to display.

So, in a five day block, a user might spend five days in the static condition seeing the same intervention each time, then five days in the rotation condition seeing randomly selected interventions each time. They are then moved into another block and the method repeats.

#### 4.5.4 Measures

We measured the *effectiveness* of the system as the number of seconds spent on the target site each day. Time, of course, does not perfectly correspond to attention or engagement behavior, as users can get distracted and not actually attend to a web page. However, prior work has generally found it to be an effective estimate (e.g., [168]). To determine whether the user is actively using a target site, we use Chrome’s internal definition

of active – the browser window and tab is focused, the computer screen is on, and there has been mouse or keyboard activity on the tab within the past minute. Time on site per day is measured as the aggregated time across sessions from midnight to midnight in the participant’s local timezone. There was one data point per user per day per targeted web site. Because time data is not normally distributed, we adopt a common practice of log-transforming the time data prior to analysis.

We measured *attrition* as the number of days the participant kept the extension enabled. We also noted if the extension was still enabled at the conclusion of our study. The browser does not send a notification to our server if a user disables the extension, so we coded instances of attrition when the server stopped receiving data from the user for over two days, with no later resumption.

As with many online field experiments, effective data cleaning is essential to accurate analysis. We excluded users who had HabitLab installed on multiple devices, to focus on site usage on a single device. We discarded days on which the target site was never visited, as in neither condition would the intervention have been shown. We also discarded the first day because participants installed the extension midway through the day, resulting in an underestimate at the day level; we likewise discarded any days on which the user uninstalled or disabled the extension, as this would again cause the measured time to be an underestimate of the actual time spent on site that day.

#### 4.5.5 Method of Analysis

For analyzing effectiveness at both the day and session level, we used a linear mixed model (LMM). We used an LMM because we have multiple samples from each user, but the number of samples from each user and in each condition is variable (because attrition may occur before they completed all conditions, or they may not visit a site on a particular day), which violates the assumptions of the repeated-measures ANOVA.

To test whether interventions decrease in effectiveness over time (H1), we focused on just data points from the static condition. The model included a term for the number of days that particular intervention had previously been seen,<sup>1</sup> a random effect for the participant, and a random effect for domain. To test linear mixed models for significance, we used a likelihood ratio test to compare a reduced model without the number of days predictor to the full model. A significant test indicates that the number of days has statistically significant explanatory power, analogous to a significant beta coefficient in a traditional regression.

To test whether static or rotated interventions increase effectiveness (H2), we used data from both the static and rotation conditions. This second LMM, predicting log time spent on the site each day, included a random effect for participant, a random effect for domain, a fixed effect for block length, and a fixed effect for condition. A likelihood ratio test compared to a reduced model without the condition variable.

To analyze whether static or rotated interventions increase attrition (H3), we used a Cox proportional hazards regression. Cox proportional hazards models predict the relative “hazard” (i.e. risk) of attrition given each predictor. This is used in the health sciences for estimating expected lifetimes when we may

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<sup>1</sup>Repeating the analysis using the number of times the intervention had been seen yields the same conclusions.

Table 4.1: Within the static condition, interventions decline in effectiveness. Longer visit lengths increase with the number of days seeing the same static intervention.

	<i>Dependent variable:</i>
	Log time spent per day
Number of days the user had seen the static intervention	0.225*
	(0.097)
(Intercept)	4.759***
	(0.392)
Observations	124

*Note:* \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

have differing durations of observations for each participant, and may have observed deaths (which correspond to attrition) for some participants but not others. Each data point consists of a point of observation, and whether the participant had experienced attrition at that point or was still active. To avoid crossing conditions in this analysis, we focus the Cox analysis on just each user's first assigned block and condition, for example a seven-day rotation block or three-day static block. Each observation consists of the length of block, and whether the user had experienced attrition by the end of the first condition for their first block. The Cox model used a single predictor: condition. The output of a Cox proportional hazards model is similar to a regression, with a significance value and estimate attached to the predictor.

#### 4.5.6 Results

In this study, participants had an average of 3.0 target sites enabled. They visited at least one target site 67% of days on average. On each of those days, participants experienced interventions an average of 3.6 times. We did not receive any feedback indicating that participants were aware of patterns in how HabitLab was rotating interventions.

##### Effectiveness of interventions over time

First we examine whether interventions decrease in effectiveness over time within the static condition. If so, rotation may be a viable strategy.

The likelihood ratio test confirms that the number of days the user had seen the static intervention affected the log of time spent on a domain per day ( $\chi^2(1) = 4.69, p < 0.05$ ), supporting H1. Each day the intervention has been previously seen increased the log time spent by 0.225 (Table 4.1). By exponentiating the log estimates, this translates into an increase of 25% on top of a baseline 117 seconds per day for each additional day the user were exposed to the static intervention.

Table 4.2: Daily time spent on sites in the static and rotation conditions. Users spend less time per day on sites in the rotation condition.

<i>Dependent variable:</i>	
Log time spent per day	
Rotation (baseline: static)	-0.417*
	(0.190)
Block length	0.018
	(0.048)
(Intercept)	4.981***
	(0.346)
Observations	370

*Note:* \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

Table 4.3: A Cox proportional hazards analysis suggests that the rotation condition substantially increases the hazard of attrition. Coefficients are log hazard ratio, so positive values indicate increased hazard and negative values indicate decreased hazard.

<i>Dependent variable:</i>	
Log hazard ratio	
Rotation (baseline: static)	0.544*
	(0.249)
Observations	217

*Note:* \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

### Effectiveness of rotation and static intervention strategies

Next, we compare whether the daily time spent on domains differs between days when participants were in the rotation and static conditions.

The likelihood ratio test found a significance difference between the full and reduced models predicting effectiveness ( $\chi^2(1) = 4.88, p < 0.01$ ), indicating that condition significantly impacted effectiveness. Relative to the static condition, rotating interventions decreased the log time spent on domains per day by 0.417 (Table 4.2), supporting H2. Exponentiating the coefficients for descriptive purposes, this translates into a shift from an estimated 146 seconds per day in the static condition to 96 seconds per day in the rotation condition, a decrease of 50 seconds (34%) per day.

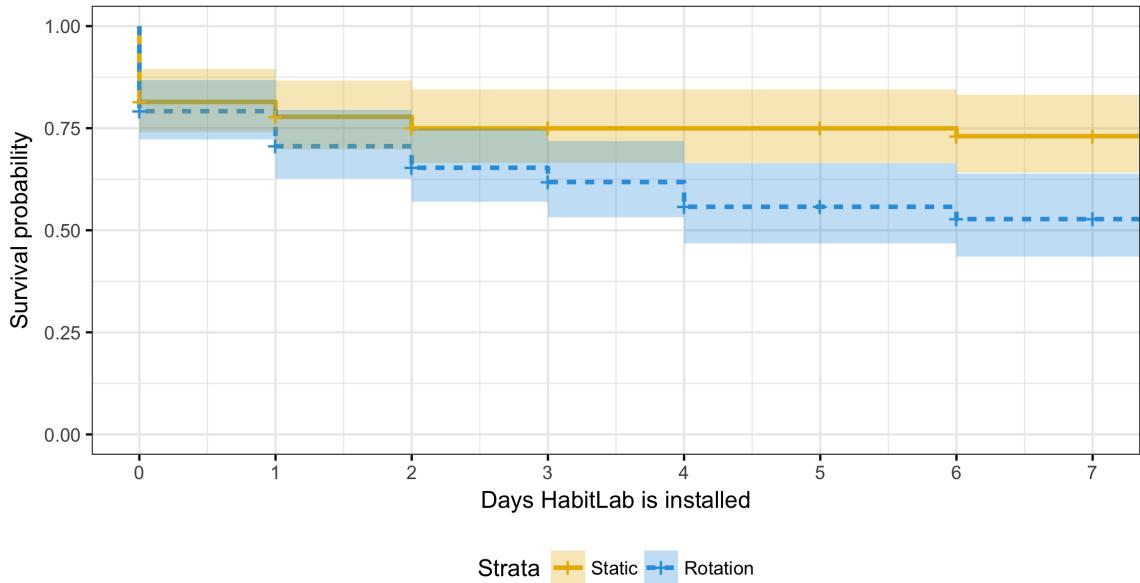


Figure 4.2: Rotating interventions increases attrition among users.

#### Attrition due to rotation and static intervention strategies

The Cox proportional hazard regression model comparing the static and rotation conditions found that attrition rates are significantly higher with the rotation condition (Figure 4.2, Table 4.3). After 2 days, 78% of users remain in the static condition, while only 71% remain in the rotation condition. After 7 days – the duration of the longest experiment block – 68% of users remain in the static condition, while only 39% of users remain in the rotation condition. These results support H3.

We considered the possibility that switching between static and rotated interventions contributes to attrition beyond simply rotating them. We analyzed this by comparing the probability of attrition on days where the condition remains the same as the previous day, to days where the condition changes – either from static to rotated, or from rotated to static. The baseline daily attrition rate is 18% when staying within the same experimental condition – 14% when staying within the static condition, and 20% when staying within the rotation condition. On the first day after switching from static interventions to rotated interventions, the attrition rate is 36% – a significant increase compared to remaining within the same condition (Fisher's exact test,  $p < 0.001$ ). However, switching from rotated interventions to static interventions does not increase the attrition rate – it remains at 18%. So we believe these effects are not due to the changes between conditions, but due to the conditions themselves – switching from static to rotated is experiencing the first instance of a rotation, and it is not surprising that the effect may be larger with the first change.

## 4.6 Study 2: Longer-Term Effects of Rotation on Attrition

Study 1 found that compared to static interventions, rotation increases effectiveness but also increases attrition. To provide additional support for our findings in Study 1 and motivate our design experiment, we present a second field study that seeks to answer the question: Does the number of interventions in the rotation affect the level of attrition? This study occurs over a longer period – ten weeks – allowing us to examine these effects in a more longitudinal setting.

### 4.6.1 Participants

Our participants were HabitLab users who installed over a 5 week period in January–February 2018 and consented to our experiment protocol. 680 users who agreed to participate. After excluding users with multiple devices, users who did not complete the onboarding process, and users who had less than two sessions on Facebook where they saw interventions – we restricted analysis in this study to users who were using Facebook because it had the most number of default interventions available – we were left with 409 participants. Demographics were similar to Study 1.

### 4.6.2 Method

This was a between-subjects study where users’ default settings for the number of enabled interventions varied depending on their condition: some users only had one default enabled intervention, and others had more. Interventions were then selected randomly from the enabled set. Among users who did not change these defaults, this enabled a between-subjects comparison of the effects of the number of interventions a user was rotating between, on retention rates.

In practice, we found that many users changed the set of interventions – 78% of participants in this study changed them over the course of using HabitLab, most often during onboarding. We wanted to retain a good user experience, but this muddied the experimental manipulation. So, we restricted analysis to the 91 users who did not change defaults. A  $\chi^2$  test found there was no significant effect of condition on whether users changed defaults ( $\chi^2(2)= 0.4671$ ,  $p=0.8$ ), suggesting that randomization remained effective even after this filter.

Unlike Study 1, this was a between subjects experiment, so there were no time blocks: participants were assigned to the condition for the duration of the study.

### 4.6.3 Conditions

Participants were randomized into three conditions. In the one intervention condition, for every site the user enabled HabitLab on, only one intervention was enabled by default. The intervention was randomly chosen among the set of default interventions for that site. This is equivalent to the static condition from Study 1.

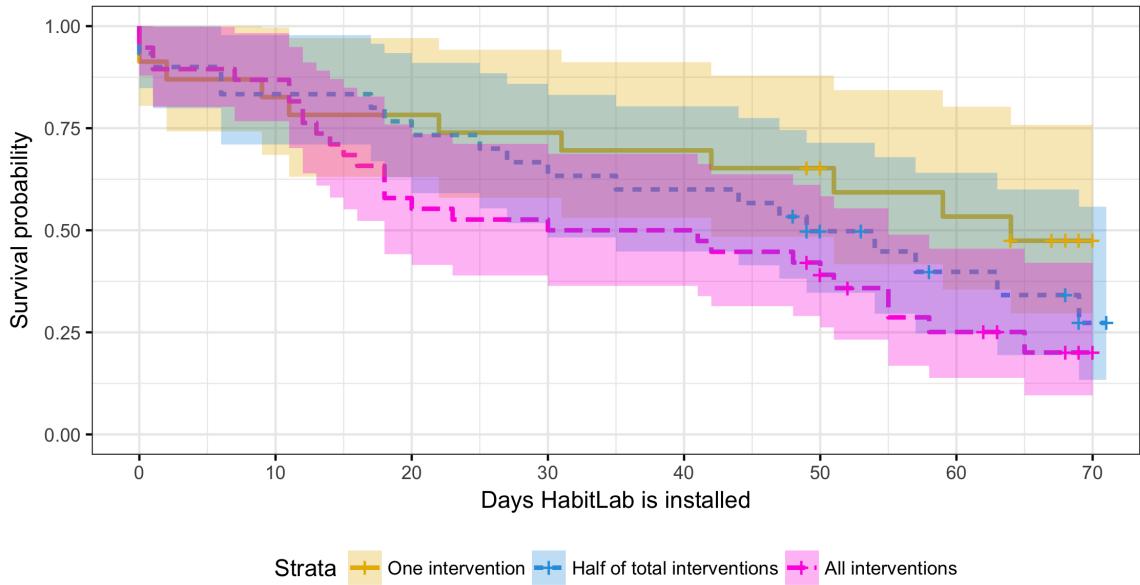


Figure 4.3: Including all interventions resulted in significantly more attrition than just one intervention.

In the all interventions condition, for every site the user enabled, all interventions that are default for that site were enabled by default. This is equivalent to the rotation condition from Study 1. In the half interventions condition, for every site the user enabled, half of all interventions that are default for that site were enabled by default. The subset was chosen randomly.

#### 4.6.4 Measures

We measured attrition, using the same procedures as those described in Study 1.

#### 4.6.5 Method of Analysis

Like Study 1, we applied a Cox proportional hazards regression model to compare attrition rates.

#### 4.6.6 Results

In this study, participants had an average of 3.3 target sites enabled. They visited at least one target site 64% of days on average. On each of those days, participants experienced interventions an average of 6.8 times.

In this longer, between-subjects experiment, attrition rates were significantly higher in the all interventions condition (Figure 4.3, Table 4.4). This agrees with the analogous result from Study 1 showing a higher attrition rate for the rotation condition. The half of total interventions survival curve falls in between that of the one intervention and all interventions conditions, but does not have a statistically significant difference.

RQ4: Are users who enable or disable interventions during onboarding less likely to attrition?

Table 4.4: A Cox proportional hazards analysis over a longer period suggests that rotating with more interventions increases the hazard of attrition.

	<i>Dependent variable:</i>
	Log hazard ratio
Half of total interventions (baseline: one intervention)	0.395 (0.380)
All interventions	0.711* (0.358)
Observations	91
<i>Note:</i>	<sup>*</sup> p<0.05; <sup>**</sup> p<0.01; <sup>***</sup> p<0.001

We compared users who enabled or disabled interventions during the first 5 minutes, compared to other users who completed the onboarding process without enabling or disabling any interventions. We found that users who enabled or disabled interventions during the first 5 minutes had a significantly lower rate of attrition, as shown in Figure 4.4.

## 4.7 Study 3: Design Interventions to Reduce Attrition

Study 1 and Study 2 collectively demonstrated that rotation increases effectiveness but also increases attrition. Why does rotation increase attrition? To understand this, we needed to understand why users uninstalled in the first place.

We performed a qualitative content analysis on the uninstall feedback left to us by users. This feedback was collected in a tab that opened automatically when users uninstalled HabitLab. The page stated that feedback would be used for research purposes. Users had the option to check boxes to agree with a set of predefined reasons they why they were uninstalling, and leave free-text feedback. We performed an inductive analysis of the free-text feedback, grouping responses by themes, reflecting on our themes, and refining our groupings until convergence.

A total of 782 users submitted the uninstall feedback form. This data represents all past users of HabitLab, and includes users outside studies 1 and 2. We use this larger dataset because only 8 participants from Study 1, and 39 from Study 2, filled out the feedback form. 751 users who submitted the form checked at least one of our predefined reasons. 274 users (36%) uninstalled because “Interventions were annoying”, 248 users (33%) uninstalled because HabitLab “Did not feel effective”, 100 users (13%) uninstalled because HabitLab “Was causing lag”, 75 (10%) uninstalled due to “Privacy concerns”, and 202 (27%) cited “Other reasons”. The total sums to more than 100% because users could check more than one reason.

A total 155 users submitted free-form textual feedback. Some users began with an incorrect mental model and uninstalled after they learned what it was doing:

- *Didn't seem what I was expected. Installed two minutes ago and removed it*

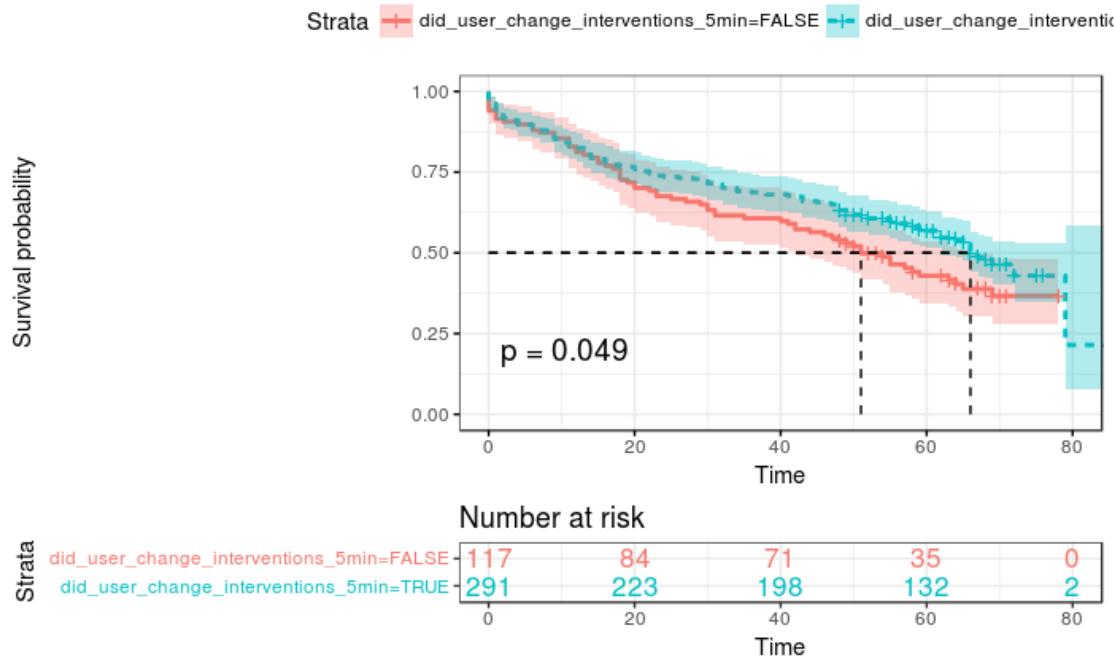


Figure 4.4: Users who change interventions during the first 5 minutes will be significantly less likely to attrition later.

- *I just didn't understand the concept before downloading*

Some users indicated they wanted more control over the intervention that was shown to them, or were simply looking for a time tracker and were not interested in interventions at all:

- *I wanted a timer for every “domain”, it can be good for statistics of time*
- *I was interested in tracking my usage to start, instead of setting interventions that I may not actually be concerned about*

Some users indicated dissatisfaction with particular interventions:

- *Mostly it was the bar covering up facebook message indicators*
- *it was just annoying you out of not using sites, not convincing you to. It became like ads, they are always there. But you don't like them and turn them off with ad-block.*

Some users wished interventions would be more forceful, or less intrusive:

- *Interventions are not forceful enough. They are too easy to click around or disable*
- *I liked the interventions but not on every page change or load, that was just a bit too much*

Finally, some users decided they simply did not want or need interventions:

- *Made me realize I don't have Facebook addiction, spending less than 30 minutes [...] per day*
- *I'm weak...*

Other themes included localization issues, performance issues, privacy concerns, accidental installations, and misattribution of other issues to HabitLab.

ome feedback indicated that users had an incorrect mental model of the system: they were expecting to see a particular intervention consistently, instead of having them rotate:

*does not fucking work, hiding the newsfeed in particular*

Other users began with an incorrect mental model and uninstalled after they learned what it was doing:

*Didn't seem what I was expected. Installed two minutes ago and removed it*

*I just didn't understand the concept before downloading and it's intentions aren't my demons as it happens*

Some users indicated they wanted more control over the intervention that was shown to them, or were simply looking for a time tracker and were not interested in interventions at all:

*I wanted a timer for every "domain", it can be good for statistics of time*

*I was interested in tracking my usage to start, instead of setting interventions that I may not actually be concerned about*

*Would have preferred a gentle log - perhaps emailed - giving usage statistics. In present form this operates like pop-up ads. Still, it was reasonably insightful into understanding my usage for the time I used it*

Some users indicated dissatisfaction with particular interventions:

*Mostly it was the bar covering up facebook message indicators*

*You covered up useful buttons. Don't do that*

*Made Facebook unusable. Which might be the point?*

Some users wished interventions would be more forceful:

*Interventions are not forceful enough. They are too easy to click around or disable*

Some users were getting fatigued from seeing too many interventions:

*Looked like some nice options for streamlining sites, but it's actually a nanny. I don't need a nanny whining at me.*

*I liked the interventions but not on every page change or load, that was just a bit too much*

*it was just annoying you out of not using sites, not convincing you to. It became like ads, they are always there. But you don't like them and turn them off with ad-block.*

Some users disliked that the intervention was reminding them that they were visiting sites:

*I noticed that I was going on imgur, youtube, facebook (my choice of addictive sites) more, after I had installed the extension. So, I'm uninstalling. I think the extension made me more conscious of the fact that I was visiting the sites, but maybe the rewards were making me go back to the site? I'm not sure*

There were also users who cited localization issues, or may not have understood the text presented during onboarding because the app was not localized to their native language:

*Translate in french please – This was from before we localized to French*

*non lo voglio – Italian for I don't want it. The extension is not localized to Italian*

Some users cited performance issues, bugs, or conflicts with other extensions:

*Was awesome, but was making chrome really slow, i mean really slow! Seems like you need to fix some memory issues*

*It's quite possible something else was causing lag – but lag was there. I also was just checking it out. I don't really use facebook or youtube*

*Catastrophic stability problems after installation; may be due to a different extension*

*Seeing if this extension is causing gmail compatibility issues*

*Wasn't sure if it is effecting battery life*

*Some users had simply been testing the app or evaluating alternatives:*

*I love the app - I'm just removing temporarily to see if it's affecting another app (Freedom.to)*

*I want to try other types of chrome extensions to block time-consuming websites and don't want to mess with your data*

*I prefer the "Forest" application*

*I forgot that I already had another program that did essentially the same thing for the computer in general instead of just for this particular browser*

*Some users had privacy concerns:*

*I was just worried, I mean it's (Anonymized) and all, but a bunch of students tracking everything I do and all my browser history, that just felt too much of a price to pay*

*Finally, many uninstalls were simply due to the extension being automatically installed on a non-work computer due to Chrome's behavior of automatically installing extensions across all devices, which is why we had restricted analyses to just users who used one device:*

*Neighbor installed this to my computer without my consent!*

*Someone else installed. Did not want it*

*Dont need it on work computer, just at home*

*this is my wasting time computer and i dont need it*

*Some users decided they simply did not want or need interventions:*

*Made me realize I don't have Facebook addiction, spending less than 30 minutes of my desktop time on it per day*

*I rarely waste time on my desktop. This would be much more useful on my mobile phone*

*I just don't use my laptop as much as I thought would be necessary for an intervention*

*Some users uninstalled due to misattributing other issues to our software:*

*For some reasons Facebook blocked me and I am trying to figure out the reasons. What I know that the Habitlab extension was deactivated (not by me) and then, boom, I was blocked*

*Finally, some users reached their goals and decided they simply didn't need the extension anymore:*

*I reached my goal to reduce time spent on certain pages. Thanks folks!*

*I used the extension to curb my Facebook habit and eventually gave up Facebook altogether - something I've been wanting to do for a long time. Thank you*

### 4.7.1 Design interventions

Based on the qualitative feedback on reasons for uninstalling, we drew on two of the most consistent themes to hypothesize why rotation may be increasing attrition:

**Hypothesis 4 (H4).** *Violation of mental model: Users may have sped through onboarding and not understood that HabitLab rotates interventions. So, when they experience a new intervention, the system violates their mental model and they disable it in confusion or frustration.*

**Hypothesis 5 (H5).** *User control: Users may be aware that the system is choosing interventions for them, but are frustrated by a lack of control over the system's behavior. They may dislike one or more of the interventions but not realize how to turn them off.*

These two hypotheses could feasibly be addressed through design interventions. Other pieces of feedback, for example how aggressive the interventions were, we judged as out of scope of the current study on rotation strategies and will pursue as future work. We developed two different interfaces, one to address mental model violation and the other to address a perceived lack of control. They are shown to users when they see a new intervention for the first time.

The first design, which we will call *mental model* (Figure 4.5), is inspired by H4: it reminds the user that HabitLab has rotated to a new intervention and gives the name of the intervention. If mental model misalignment was the issue, this design might help explain to the user what the system is doing and why. The second design, which we will call *user control* (Figure 4.6), is inspired by H5: it includes the message in the information design but also adds a toggle option to allow the user to turn off the new intervention for future visits without needing to visit HabitLab's settings. If lack of control was the issue, this design may give sufficient control so that users keep HabitLab enabled.

e hypothesized:

**Hypothesis 6 (H6).** *The information and control designs will have less attrition than a control design.*

### 4.7.2 Experiment Design

We ran a between-subjects design where we randomized the design shown to new users of HabitLab and tested whether it impacted attrition over a period of one week, similar to Study 1.

### 4.7.3 Participants

Our participants were HabitLab users who installed over a 10 day period in April 2018. There were a total of 282 users who installed and agreed to participate. We removed users who were not new users (e.g. an existing user installing on a new device, or a former user reinstalling the system), and users who left before they saw their first intervention. This leaves us with data from 93 participants. Demographics, estimated by Google Analytics, were similar to Study 1.

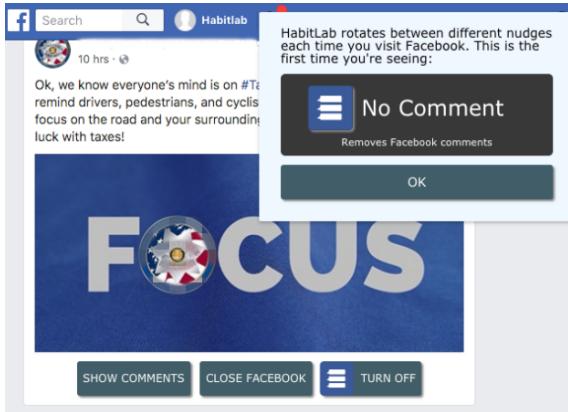


Figure 4.5: Mental model interface: each time the user sees a new intervention, HabitLab names it and explains about rotation.

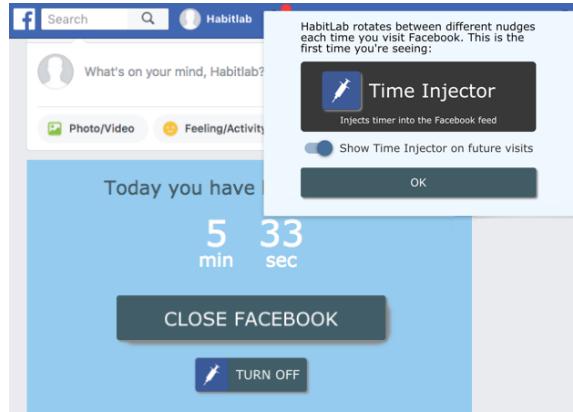


Figure 4.6: User control interface: in addition to the mental model information, HabitLab gives users a direct interface to disable the new intervention.

#### 4.7.4 Method

Participants installed HabitLab and set it up as described in the Study 1 and Study 2. They used HabitLab in the course of their normal web browsing activity. HabitLab rotated between randomly chosen interventions on each visit to the chosen web page for all users, equivalent to the rotation condition in Study 1. Each time the user experienced a new intervention that they had not seen before, however, HabitLab might show an explanation design in the corner of the browser.

#### 4.7.5 Conditions

There were three conditions for this study. In the no design condition, users saw no message, equivalent to the rotation condition from Study 1. In the mental model condition, users were shown the informational intervention (Figure 4.5) to remind them that the system rotates interventions. In the user control condition, users were additionally given control over whether to turn off each new intervention without needing to visit the settings screen (Figure 4.6).

#### 4.7.6 Measures

Our main dependent variable was attrition—how many days users kept the system installed by the end of the study, seven days after installation. The measure of attrition was the same as in Study 1 and Study 2. We also measured effectiveness, using the same method as Study 1.

#### 4.7.7 Method of Analysis

To analyze attrition, we again used a Cox proportional hazards regression model, similar to Study 1, using interaction design as the predictor variable. To analyze effectiveness, we used a LMM predicting log time on

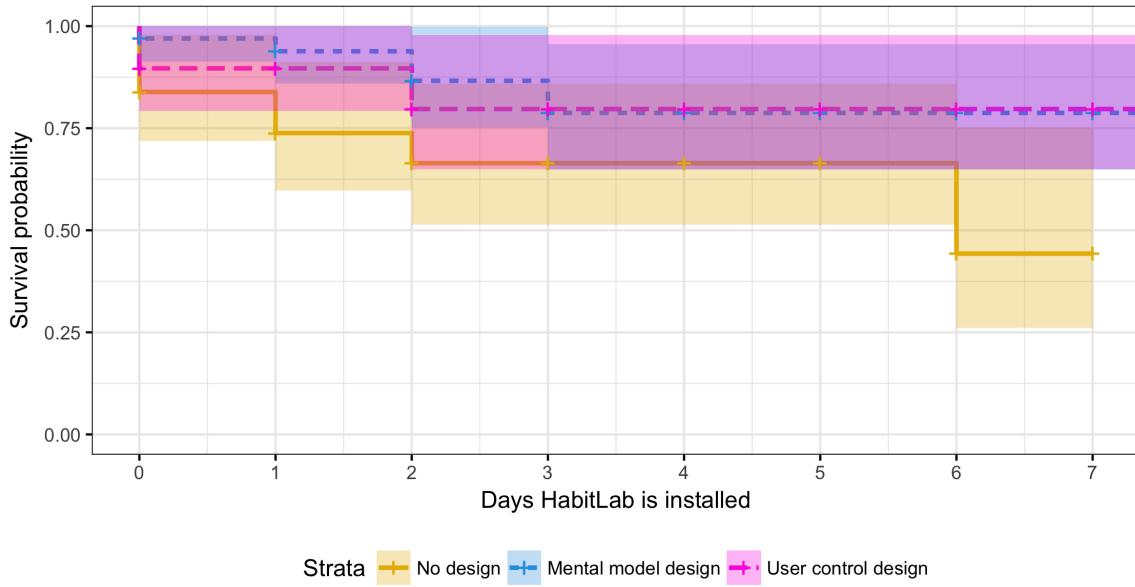


Figure 4.7: Reminding users about how the rotations worked every time a new intervention was introduced significantly reduced attrition rates.

site per day, with a fixed effect for condition, and random effects for participant and domain. Data cleaning followed the same procedures as Study 1.

#### 4.7.8 Results

In this study, participants had an average of 2.9 target sites enabled. They visited at least one target site 71% of days on average. On each of those days, participants experienced interventions an average of 6.6 times.

The Cox proportional hazard regression indicates that the mental model design significantly reduces attrition rates relative to no design ( $p < 0.05$ , Figure 4.7, Table 4.5). This result supports H4. After seven days, 79% of participants in the mental model condition remain, while 80% remain in the user control condition and only 44% remain in the no design (control) condition. In other words, the intervention conditions more than halved the attrition rate, from 56% to 21% attrition. Adding the additional option to permanently turn off the intervention the first time it is seen is not significantly different from no design given our sample size.

There was no effect of condition on effectiveness: the full model was not significantly more explanatory than the reduced model without the condition variable ( $\chi^2(2) = 1.46$ , n.s.). So, these interventions did not reduce effectiveness while they were improving attrition.

Table 4.5: A Cox proportional hazards analysis suggests that the informational intervention that corrected users' mental models was successful in reducing attrition due to rotation. Coefficients are log hazard ratio.

<i>Dependent variable:</i>	
	Log hazard ratio
Mental model design	−1.015* (0.494)
User control design	−0.869 (0.527)
Observations	93

*Note:* \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

## 4.8 Discussion

Our findings suggest that changing behavioral interventions can be beneficial from the perspective of efficacy, but detrimental to retention. By showing simple messages when presenting new interventions, we can improve users' mental models, and reduce attrition from changing interventions.

In addition to the interface-based techniques we have presented to combat detrimental effects of changing interventions, algorithmic techniques can also help. For example, in the context of a multi-armed bandit, potential algorithmic techniques include:

1. Limiting the exploration speed such that users are not overwhelmed by the rate at which they are seeing new interventions.
2. Modeling individual interventions' likelihood of attrition, and favoring algorithms which are less likely to cause attrition if needed to keep the user around longer.

There are also additional interface-based techniques that may be helpful in reducing attrition from changing interventions, but that we have not tested, such as:

1. Making how new interventions are introduced predictable and known to the user.
2. Allowing users a choice of intervention when we introduce new interventions.

### 4.8.1 Limitations

This work featured deployment periods of a few weeks. This may not be enough time to observe some very long-term effects: for example, some changes in intervention effectiveness set in only after months or years [91]. That said, given the fast turnover rate which is observed with behavior-change software, even short-term effects of changing interventions on attrition can be important.

While we believe our general finding about the double-edged nature of changing interventions may apply to other behavior-change contexts, particular parameters—such as speed at which users grow blind to an intervention, may be domain-specific.

One shortcoming of our Study 1 design is that we cannot rule out the possibility that our observed increase in effectiveness is due to selective attrition, rather than due to benefits from the rotation. Namely, it is possible that observing rotation may selectively lead to uninstallation for users for whom interventions are ineffective. To rule out this possibility, we will need to investigate ways to maintain retention in the presence of rotation, and see whether the improvement in effectiveness relative to a static intervention still remains. It may also be possible to design intention-to-treat analyses that discount attrition in measures of effectiveness. Furthermore, while we observed that the first visit is longer than subsequent visits when users visit sites multiple times per day, but this effect may be due to temporal usage patterns rather than intervention effectiveness.

Because users have differing preferences, interventions may have differing rates of attrition for each user. An ideal retention-maximizing system would not assign interventions randomly, but would personalize interventions to each user. Assuming there is a novelty component to attrition — i.e., users quit because they grow bored of the same intervention — then a system which intelligently times interventions to minimize attrition can in theory have lower attrition than even the best static intervention. There are 2 difficulties in making this a reality: first is needing to learn to correctly predict which intervention would minimize attrition for a user at a given time, a reinforcement learning problem. Second, as shown by the increase in attrition when using a naïve rotation strategy, a system that switches between interventions also needs to overcome the barriers of needing users to develop more complex mental models, and ensuring that users feel in control.

#### 4.8.2 Design reflections on social computing and behavior change

Social systems are inherently tied to behavior change and retention. Social networks and other social apps and services make heavy use of gamification and behavior change techniques to drive engagement and boost retention [50, 28]. A system like HabitLab that helps users use these services less thus occupies an interesting space: it is modifying the service to hide the features that serve to boost engagement, helping users break away from their addiction to the site.

But we tread a fine line: behavior change systems themselves suffer from attrition, so we may sometimes need to make tradeoffs between better retaining users and helping them regulate their behaviors. For example, the Facebook interventions in HabitLab with the lowest attrition—those that passively show time spent—are among the least effective. Is telling users that the system is helping them more than it actually is a form of benevolent deception [4] that would ultimately help boost retention and help users achieve their goals? Would gamifying the system with social features, making users connect with friends and keep tabs on their friends' social media usage, help boost retention and effectiveness—even though users may lose time engaging with social features?

We believe that novelty is an underlying mechanism for the improvement in effectiveness we observed when interventions are rotated. This leads us to speculate: would it be an effective and practical strategy to scale up the number of interventions, so that we are rotating between interventions from a huge pool of hundreds of interventions? Or do the improvements in effectiveness that we can expect from rotating increasing numbers of interventions have limits? We speculate that increasing the number of interventions in rotation will have high initial benefits for the first few additional interventions, but will have declining benefits as more interventions are added, as the probability of repeatedly seeing a recently-seen intervention grows increasingly small.

## 4.9 Conclusion

Behavior change intervention effectiveness declines as interventions are repeatedly shown to the user. This decline can be combated by rotating between a stable set of different interventions. Rotating interventions increases attrition, but user interface changes can ameliorate the issue. Taken together, these results suggest opportunities to build behavior change systems that operate more like coaches and tutors: they might explore different strategies to find what works well, and then occasionally rotate to keep things fresh.

## Chapter 5

# Do Productivity Interventions Save Time or Just Redistribute It?

### 5.1 Introduction

We use productivity behavior change interventions to try to keep ourselves in focus. But do these systems truly save us time? Or do they just redistribute the time elsewhere? In other behavior change domains, interventions sometimes have effects on behaviors other than the ones they were targeting [151, 36].

One possibility is that interventions narrowly impact just the goal that they target, and have no effect on time spent elsewhere. We will refer to this as the *isolated effects hypothesis*. Taking the relationship between time spent on Facebook and Instagram as an example, the isolated effects hypothesis would predict that an intervention that helps reduce time on Facebook should have no effect on time spent on Instagram. Persuasive systems often claim to result in the intended behavioral changes without observable consequences elsewhere, lending support for this hypothesis [16, 38, 100, 11, 12]. If the isolated effects hypothesis is true, overall productivity can be boosted through interventions that individually target each goal.

However, people have a limited supply of willpower [18], can maintain focus for only so long [72, 40, 105], and need downtime — so perhaps the time saved is actually just redistributed to other unproductive applications. We will refer to this as the *redistribution hypothesis*: saving time on one unproductive application results in an increase in time spent on other unproductive applications. Returning to our example of a productivity intervention targeting Facebook, redistribution would hypothesize that an intervention that reduces time on Facebook will increase time spent on Instagram. Redistribution may be partial, where the time redistributed is some fraction of what was saved. Or more bleakly, redistribution may be total, where the time redistributed is entirely shifted to other applications and there is no overall improvement in productivity.

A third possibility is that saving time on one application breaks a habit loop [46] and reduces time

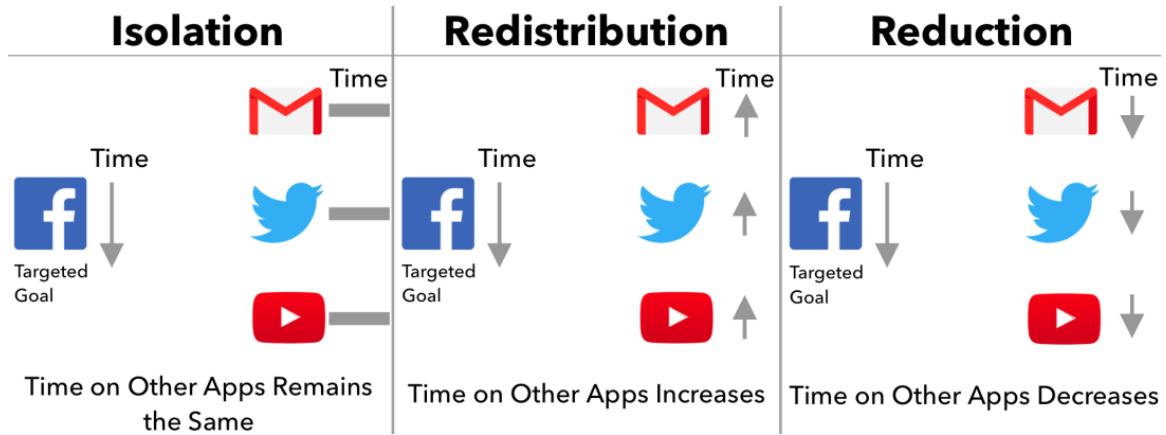


Figure 5.1: When interventions reduce time on a targeted goal such as Facebook, the time saved may (left) be isolated from effects on other goals, (center) be redistributed to other goals, or (right) decrease time spent on other goals.

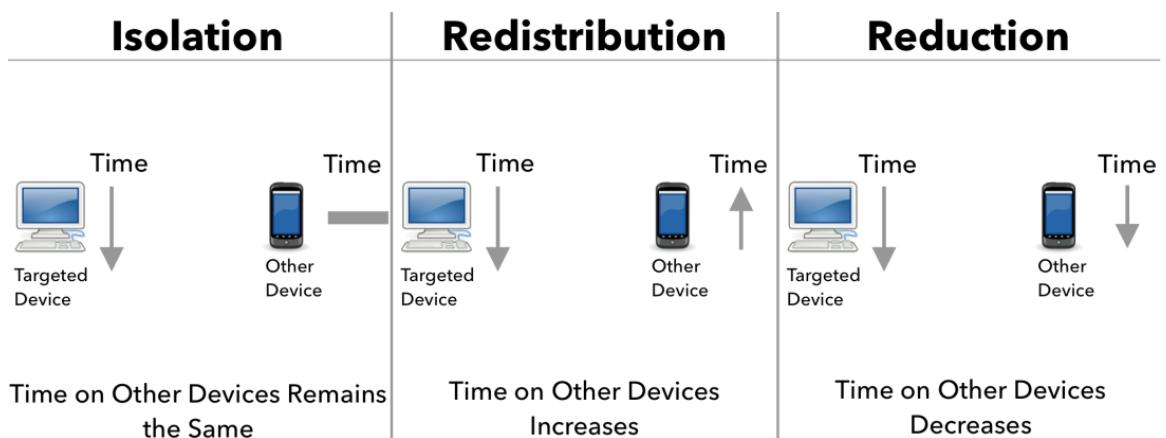


Figure 5.2: When interventions reduce time on a targeted device e.g. a browser, the time saved may (left) be isolated from effects on other devices, (center) be redistributed to other devices, or (right) decrease time spent on other devices.

spent on other applications as well, so the actual net improvement in productivity is even better than just what is saved on the target application. We will refer to this as the *reduction hypothesis*. Returning to our example of a productivity intervention targeting Facebook, this would hypothesize that an intervention that reduces time on Facebook will also reduce time on Instagram. Perhaps once we enter “procrastination mode” and visit one unproductive application, we wind up chaining together visits to another unproductive application, and another—but if a productivity intervention helps us break the chain early on, we will never visit the later unproductive applications.

These three hypotheses lay out the three possibilities of what happens to other goals when we intervene on a focal goal (Figures 5.1–5.2): time on those other goals might stay the same (isolated effects), go up (redistribution), or go down (reduction). In this chapter, we seek to adjudicate between these hypotheses using HabitLab [90], an in-the-wild productivity experimentation environment that users can voluntarily participate in by installing. Prior work described HabitLab as a Chrome browser extension; in this chapter we created and deployed a companion HabitLab Android application, allowing us to study any redistribution of time that might be happening across devices, as when a user avoids Facebook on their browser but ends up checking Facebook on their phone instead.

After installing and agreeing to our experimental protocol, users specify what they wish to reduce time on, which we term *goals*. In the case of the Android version, goals take the form of applications (apps), whereas on the Chrome extension goals are sites. We then deploy interventions to help users reduce their time on these goals, which can appear when the user visits a website (Chrome) or app (Android). To study redistribution, we periodically manipulate the frequency at which interventions appear for each goal – if the goal is in the frequent condition that week, it will appear every time the user visits that application, whereas if the goal is in the infrequent condition that week, it will appear on 20% of visits. This experimental design allows us to observe the effects of a goal being in the *frequent* setting not only on how much time users spend on that focal goal, but also what happens to time on other goals when that focal goal is in the frequent setting.

Our analysis first begins by seeing whether interventions are effective at reducing time on the focal goal, disregarding any possible redistribution effects. We do so by comparing time spent per day on the application on weeks where interventions are shown frequently, vs those weeks where interventions are shown infrequently. We find that they are effective, with time spent on goal sites reduced by 8.0% on the Chrome version, and time spent on goal apps reduced by 37.3% on the Android version.

Next, we investigate whether time is redistributed to other sites/apps on the same platform (browser or mobile) when interventions are frequently shown. We find that giving interventions within the browser produces a reduction effect, with users using sites/apps less when there are more interventions shown on other sites/apps – however, effects of interventions are isolated on mobile.

Finally, we investigate whether time is redistributed across devices. We do not observe any significant time redistribution effects in either direction.

This chapter contributes a look into potential unintended side effects of productivity interventions on

other sites, apps, and devices. We find that productivity interventions do not appear to have deleterious second-order effects on goals other than the ones they are targeting, and in some cases, may even have beneficial second-order effects by breaking habit loops.

## 5.2 Related Work

Measuring the effectiveness of a persuasive system remains a major challenge in the design of behavior change systems. While behavior change systems can be effective during experiments [16, 39, 164], many review papers are more restrained in whether behavior change systems remain effective outside studies and bring longitudinal behavioral change [111, 22, 114, 61]. Because behavior changes are long and complex processes, the efficacy of a persuasive system is often difficult to measure [125]. For instance, an intervention promoting healthy habits, which was effective in changing participants' eating habits, might reduce their physical activities, which were not measured in the experiment [36]. Likewise, a system promoting increased physical activity may be unable to observe effects on participants' eating habits [43]. Compared to prior work, our study examines these spillover effects in the context of a more complete ecosystem, including both desktop browsers and mobile devices.

### 5.2.1 Multitasking and Cyberslacking

Cyberslacking, referred to as non-work-related computing, is the use of Internet and mobile technology during work hours for personal purposes [163, 121, 76, 95]. One study found that employees spent at least one hour on non-work-related activities during a regular work day [163]. Researchers also reported that non-work-related Internet usage comprises approximately 30%–50% of total usage [1, 70].

Unproductive time begets further unproductive time. For example, increased time spent online can increase sleep debt, which in turn leads to more time spent online [106]. Likewise, the Hook Model claims that many of the most addictive online sites use a cycle of investment techniques to keep users coming back—for example, making a post on Facebook may result in future notifications, which will in turn will get the user to come back and make more posts [50]. Finally, sites such as Facebook, Reddit, Twitter, and Buzzfeed are filled with links to each others' content, so it may be the case that increasing usage of one will increase usage of others. If productivity interventions are able to break this vicious cycle of procrastination for one application, they may actually reduce time spent on other unproductive applications as well.

### 5.2.2 Distribution Of Unproductive Time

In this section, we will examine related studies in behavior change systems to develop testable hypotheses regarding the research question.

Multitasking has become ubiquitous in today's workplaces [15, 103, 25]. Multitasking is both essential and unavoidable in the workplace [58, 104], and it takes 11 minutes on average before people switch to a new task [40].

Studying behavior change effects across multiple devices is important: focusing on a single platform will myopically miss unproductive behaviors on other platforms. Attention is fragmented in both mobile and traditional desktop environments [93, 103]. The time spent on mobile devices has increased more rapidly than time on computers or TVs [19, 33]. On the other hand, mobile applications have been regarded as substitutions of websites in many studies [159]. Large technology companies such as Facebook and Amazon have been focusing on user growth on mobile devices [93].

However, interventions may result in unintended outcomes [63, 61, 152]. Specifically, while some interventions may be highly effective at achieving the measured goal of a behavioral change system, they may reduce desired outcomes elsewhere [61]. In one health-related intervention, while the physical activity of participants increased, calorie intake also increased, working against the goal of promoting a healthy lifestyle [23]. Similarly, using peer pressure to build confidence for students at school would, in turn, lower their self-esteem which actually was opposite to the goal of augmenting confidence [152].

### 5.3 Research Questions

Unproductive time begets further unproductive time. For example, increased time spent online can increase sleep debt, which in turn leads to more time spent online [106]. Likewise, the Hook Model claims that many of the most addictive online sites use a cycle of investment techniques to keep users coming back—for example, making a post on Facebook may result in future notifications, which will in turn will get the user to come back and make more posts [50]. Finally, sites such as Facebook, Reddit, Twitter, and Buzzfeed are filled with links to each others' content, so it may be the case that increasing usage of one will increase usage of others. If productivity interventions are able to break this vicious cycle of procrastination for one application, they may actually reduce time spent on other unproductive applications as well.

The importance of understanding the effectiveness of productivity interventions in a complete ecosystem and the rising awareness of unproductive time spent on mobile devices call into focus: would productivity interventions reduce net unproductive time? Or is it a weak palliative with little discernible effect? This led to our research question:

**Research Question (RQ).** *Do productivity interventions reduce net unproductive time, or just redistribute it to other applications, sites, and devices?*

Studying behavior change effects across multiple devices is important: focusing on a single platform will myopically miss unproductive behaviors on other platforms. Attention is fragmented in both mobile and traditional desktop environments [93, 103]. The time spent on mobile devices has increased more rapidly than time on computers or TVs [19, 33]. On the other hand, mobile applications have been regarded as substitutions of websites in many studies [159]. Large technology companies such as Facebook and Amazon have been focusing on user growth on mobile devices [93].

However, interventions may result in unintended outcomes [63, 61, 152]. Specifically, while some interventions may be highly effective at achieving the measured goal of a behavioral change system, they

may reduce desired outcomes elsewhere [61]. In one health-related intervention, while the physical activity of participants increased, calorie intake also increased, working against the goal of promoting a healthy lifestyle [23]. Similarly, using peer pressure to build confidence for students at school would, in turn, lower their self-esteem which actually was opposite to the goal of augmenting confidence [152].

In our system, the time spent on unproductive activities might be decreased in one application yet increased in others. These prompt our hypotheses:

**Hypothesis 7 (H7).** *Within a single device, productivity interventions will cause the time spent on targeted sites and apps to be redistributed to other sites and apps.*

**Hypothesis 8 (H8).** *Between computers and mobile devices, productivity interventions will cause the time spent on one device to be redistributed to other devices.*

## 5.4 Experiment Platform: HabitLab

We conducted the studies in this chapter using the browser and android versions of HabitLab. At the time the studies presented in this chapter were conducted, the Chrome version had over 8000 daily active users, and the Android version had over 500 daily active users. The list of interventions that were included in HabitLab at the time of this study is included at the end of this chapter.

## 5.5 Study: Redistribution of Time Within and Across Devices

In this study we aim to analyze whether productivity interventions are reducing or redistributing time. We pursue this through an experiment and three sets of analyses: (1) *Within-device redistribution of time, in the browser*. For example, this would be the effects on time spent on non-Facebook websites, due to interventions that run when visiting the Facebook website. (2) *Within-device redistribution of time, on mobile devices*. For example, this would be the effects of time spent on non-Facebook applications, due to interventions that run when using the Facebook app. (3) *Cross-device redistribution of time*. For example, this would be the effects of time spent on Facebook on the phone, due to interventions that run when visiting the Facebook website.

### 5.5.1 Participants

Participants in this study consisted of new HabitLab users who installed either the HabitLab Chrome extension or Android app over a period of 132 days (approximately 19 weeks) in July through December 2018. 3747 users installed the HabitLab Chrome version over the course of our experiment and consented to our research protocol. 1483 users did so for the Android version. 298 installed both and signed in with their Google accounts, allowing us to analyze their usage across devices. We discarded participants who were not new users of HabitLab, since some users were re-installs or new devices for existing users. We also

Table 5.1: Data Summary. Note that the duration of 132 days are users who kept it installed the longest, but as users can freely install/uninstall we do not have 132 days of data on all users.

	<b>Browser</b>	<b>Android</b>	<b>Synced</b>
<b>Time Duration</b>	132 days	132 days	132 days
<b>No. of Users</b>	1790	782	82
<b>No. of Sessions</b>	4.8 million	11.3 million	3.8 million

discarded participants who did not complete the onboarding process, or who uninstalled the system before they saw their first intervention. This left us with 1790 participants for Chrome, 782 participants for Android, and 82 participants for whom we could analyze usage across both. A summary of our dataset is shown in Table 5.1.

### 5.5.2 Method

In order to observe time redistribution effects between a focal goal and other goals due to interventions, we would ideally randomly turn interventions on and off for goals, then observe the effects on other goals. However, because HabitLab informs users that it will show interventions on goals that they select, there would be negative consequences (e.g., user confusion and dissatisfaction) if interventions for an application disappeared entirely for a week. Therefore, we opt to vary frequency rather than entirely turn off interventions for a goal each week.

So, for each goal on each device, we randomize frequency of interventions each week. On weeks where a goal is set as frequent, an intervention is shown on every visit to the app or site. On weeks where a goal is set as infrequent, an intervention is shown with probability 0.2 on every visit to the app or site. We choose this methodology of varying frequency to approximate the effects of turning interventions entirely on or off.

We analyze the effects interventions have on overall time spent on goals in the browser and mobile environments. We do so with a linear mixed model, which models the relationship between a dependent variable of time spent that day on a goal, an independent variable of goal frequency (frequent or infrequent), and categorical variables for the user and the goal site or app (e.g., Facebook, YouTube, Reddit) as random effects. We run the model separately on both the data from the browser and mobile versions. Our results here can also be replicated with a simpler model of an independent sample t-test modeling the effects of frequency on time spent.

### 5.5.3 Intensity

Frequency measures how much a user is being nudged in a single goal, but our experiment also needs to measure how much a user is being nudged overall, across all goals on the platform. This allows us to, for example, measure whether mobile device usage increases when browser interventions are overall

more frequent, or whether time spent on non-goal sites increases when interventions are more frequent on goal sites. So, we define a measure of *intensity*: the percentage of sessions on any goal that triggered an intervention. For example, if the goal apps are Facebook and YouTube, the user visited Facebook 10 times and saw interventions 2 times, and visited YouTube 3 times and saw interventions 3 times, then the intensity is  $\frac{5}{13} = .38$ . Intensity will naturally vary over time as goals are re-randomized into *frequent* and *infrequent* conditions, with more frequent goals increasing intensity and more infrequent goals decreasing intensity. This randomization occurs for all goals simultaneously, once a week. We chose this intensity metric for our analysis, as opposed to alternatives such as raw number of times interventions were seen, because: 1) it is independent of the dependent variable, total time spent; 2) it is independent of the number of times the user visits a site/app; 3) it is guaranteed to be between 0 to 1, which is useful for interpretation; and 4) it can be used for both within-device and cross-device analysis.

For each goal, we also define a measure of *intensity of other goals*. This is the intensity measure excluding the current goal. We will use it for analyzing redistribution of time within device: when intensity of other goals varies, what is the effect on time spent on a target goal?

#### 5.5.4 Time Redistribution

##### Within Device

We analyze the effects of interventions on time redistribution within device. We define *time redistribution within device* as an increase in time spent on the goal on the device, as a result of a change in intensity of other goals. For example, an increase in time spent on YouTube as a result of turning Facebook interventions on would be an example of time redistribution from Facebook to YouTube.

We do so with a linear mixed model, which models the relationship between a dependent variable of time spent that day on all goals, an independent variable of intensity of goals, as well as the user as a random effect. We run the model separately on both the data from the browser and mobile versions. Because our time data is log-normally distributed, we fit our linear mixed models to log time.

##### Across Device

We analogously define time redistribution between devices as an increase in time spent on the other device, as a result of interventions increasing in frequency in the other device. For example, an increase in time spent on Facebook on the browser, as a result of increasing the frequency of interventions on mobile would be an example of time being redistributed from mobile to browser.

We do so with a linear mixed model, which models the relationship between a dependent variable of time spent that day on all goals on one device, an independent variable of intensity of goals on the other device, and the user as a random effect. We run the model separately on data in both directions: one analyzing the effects of browser intensity on time spent on mobile, and another analyzing the effects of mobile intensity on time spent on the browser. We again log transform our time data for analysis.

Table 5.2: Browser: Frequent interventions for a goal site cause a reduction of time spent on the site.

<i>Dependent variable:</i>	
Log daily time on site	
Frequent (1=true)	-0.085*** (0.010)
Baseline	5.904*** (0.224)
Observations	96,489

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

## 5.6 Results

First, we establish that our interventions are effective – that is, increasing the frequency of intervention on a goal app reduces time on that app. Next, we confirm that increasing intensity on a device reduces time on goal apps on that device. Then, we analyze redistribution effects within device – that is, whether increasing intensity effects time on non-goal apps. We also analyze redistribution effects across devices – that is, whether increasing intensity on one device effects time on goal apps on the other device. Finally, we build intuition for the underlying mechanisms by exploring what happens after users visit goal applications.

### 5.6.1 Are interventions effective?

#### Browser

**Yes.** We look at the effect of frequency of interventions on time spent on a day on a site, controlling for the user and the goal. We find a significant reduction in time spent on day on an app, when interventions for that goal are frequently shown that day ( $p < 0.001$ ), as shown in Table 5.2. Estimated log time on a goal when infrequent is 5.747 (313 seconds), while for frequent goals this is reduced to 5.665 (288 seconds). Hence, our methodology of increasing intervention frequency is effective at reducing time on sites.

#### Mobile

**Yes.** We look at the effect of frequency of interventions on time spent on a day on an app, controlling for the user and the goal. We find a significant reduction in time spent on day on an app, when interventions for that goal are frequently shown that day ( $p < 0.001$ ), as shown in Table 5.3. Estimated log time on a goal when infrequent is 5.928 (375 seconds), while for frequent goals this is reduced to 5.462 (235 seconds). Hence, our methodology of increasing intervention frequency is effective at reducing time on apps.

Table 5.3: Mobile: Frequent interventions for a goal app cause a reduction of time spent on the app.

<i>Dependent variable:</i>	
Log daily time on app	
Frequent (1=true)	-0.045*** (0.011)
Baseline	5.254*** (0.057)
Observations	96,147

Note: \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

Table 5.4: Browser: Increasing intensity results in a reduction of time spent each day on all goal domains

<i>Dependent variable:</i>	
Log daily time spent on all goal sites	
Browser Intensity	-0.187*** (0.016)
Baseline	6.929*** (0.033)
Observations	57,204

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

### 5.6.2 Is time spent on goals reduced when there is higher intensity?

In the previous analysis we have shown that increasing frequency of interventions on a single goal allows us to observe reductions in time spent on that goal, on both the browser and mobile platforms. In this section we will show that increasing intensity also allows us to observe reductions in total time spent on all goal apps, on both platforms. This allows us to confirm the validity of our intensity metric, as well as allow us to analyze the aggregate usage of all goal apps on each device. This will be necessary for our later analyses of redistribution effects within device as well as between devices.

#### Browser

**Yes.** We look at the effect of intensity on total time spent on goal sites each day, controlling for the user. We find a significant reduction in total time spent on goal sites when intensity is higher than day ( $p < 0.001$ ), as shown in Table 5.4. Estimated log total time on goal sites with low intensity is 6.885 (978 seconds), while with high intensity this is reduced to 6.758 (861 seconds). Hence, when interventions are more frequent in aggregate on the browser (which intensity captures), overall time on goal sites is reduced.

Table 5.5: Mobile: Increasing intensity results in a reduction of time spent each day on all goal apps

<i>Dependent variable:</i>	
Log daily time spent on all goal apps	
Mobile Intensity	-0.049* (0.025)
Baseline	8.300*** (0.042)
Observations	22,970

Note: \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

Table 5.6: Browser: Increasing intensity results in a reduction of time spent each day on non-goal sites

<i>Dependent variable:</i>	
Log daily time spent on all non-goal sites	
Browser Intensity	-0.169*** (0.016)
Baseline	8.207*** (0.028)
Observations	57,204

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

## Mobile

**Yes.** Like the browser, we look at the relationship between increasing intensity on one's mobile phone and the total time spent that day on one's goal applications. We find a significant decrease ( $p < .05$ ) in goal time spent, as shown in Table 5.5. Estimated log total time on goal apps with low intensity is 8.146 (3450 seconds), while with high intensity this is reduced to 8.031 (3075 seconds). Hence, when interventions are more frequent in aggregate on mobile (which intensity captures), overall time on goal apps is reduced.

### 5.6.3 What is the effect of increasing intensity on other, non-goal apps and sites?

#### Browser

**Reduction.** We look at the effect of intensity on total time spent on non-goal sites each day, controlling for the user. We find a significant reduction in total time spent on non-goal sites when intensity is higher than day ( $p < 0.000005$ ), as shown in Table 5.6. Estimated log total time on non-goal sites when intensity=0 is 8.207 (3667 seconds), while when intensity=1 this is reduced to 8.038 (3096 seconds). This is the effect predicted by our global reduction hypothesis.

Table 5.7: Mobile: Increasing intensity has no significant effect of time spent on non-goal apps.

<i>Dependent variable:</i>	
Log daily time spent on non-goal apps	
Mobile Intensity	0.035 (0.020)
Baseline	9.277*** (0.044)
Observations	22,970

Note: \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

Table 5.8: Mobile: Varying intervention intensity has no effect on total time spent on browser goal sites

<i>Dependent variable:</i>	
Log daily time spent on browser goals	
Mobile Intensity	0.045 (0.218)
Baseline	6.736*** (0.251)
Observations	1,312

Note: \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

## Mobile

**No effect (isolation).** We do not observe a significant effect of Android intensity on time outside of goals, as shown in 5.7. This suggests that reducing time within Android is an “isolated” behavior. Note there is an insignificant trend towards increasing time on non-goal sites with increasing intensity ( $p = 0.07$ ).

### 5.6.4 Is time redistributed between devices?

#### Mobile to Browser

**No effect (isolation).** We look at the effect of mobile intervention intensity, on total time spent on browser. We find no significant effect ( $p > .5$ ), as shown in Table 5.8.

#### Browser to Mobile

**No effect (isolation).** We look at the effect of browser intervention intensity, on total time spent on mobile. We find no significant effect ( $p > .5$ ), as shown in Table 5.9.

Table 5.9: Browser: Varying intervention intensity has no effect on total time spent on mobile goal apps

<i>Dependent variable:</i>	
Log daily time spent on mobile goals	
Browser Intensity	0.064 (0.068)
Constant	8.219*** (0.128)
Observations	1,312

Note: \*p<0.05; \*\*p<0.01; \*\*\*p<0.001

### 5.6.5 Destination tracking

Finally, to build intuition as to the mechanism by which the above effects are happening, we analyzed what happens after users open their goal applications. We visualized the flow of sessions from the 10 most widely chosen goal apps and sites in our dataset as Sankey diagrams (Figures 5.3 and 5.4). On mobile, a majority of sessions end up going to another application, followed by turning off the phone, as shown in Figure 5.3. On browsers, the majority of sessions went to other sites, as shown in Figure 5.4. We can also observe differences in goals users choose on mobile as opposed to desktop – on mobile, the most popular apps tend to be messaging apps, whereas on the browser they tend to be content aggregators.

## 5.7 Limitations

Our methodology varied frequency of interventions, instead of comparing having interventions completely on vs completely off. This approach reduces the size of effects we can observe compared to having interventions completely on or completely off. Our approach is also sensitive to variance in the effectiveness levels of the interventions. Some interventions may be more aggressive than others and change users' behavior more drastically even with low frequency. This difference may alter time re-distributions due to varied frequency.

We did not measure time spent on platforms that HabitLab does not support. For instance, HabitLab users may use Facebook on tablet devices, watch TV or engage in other activities that are considered unproductive aside from browsing on a desktop or on an Android phone. These behaviors may potentially change how time redistributed, but we are unable to track it.

Additionally, our study explores time redistribution in the context of productivity. It is possible that this context may not generalize to other behavior change regimes.

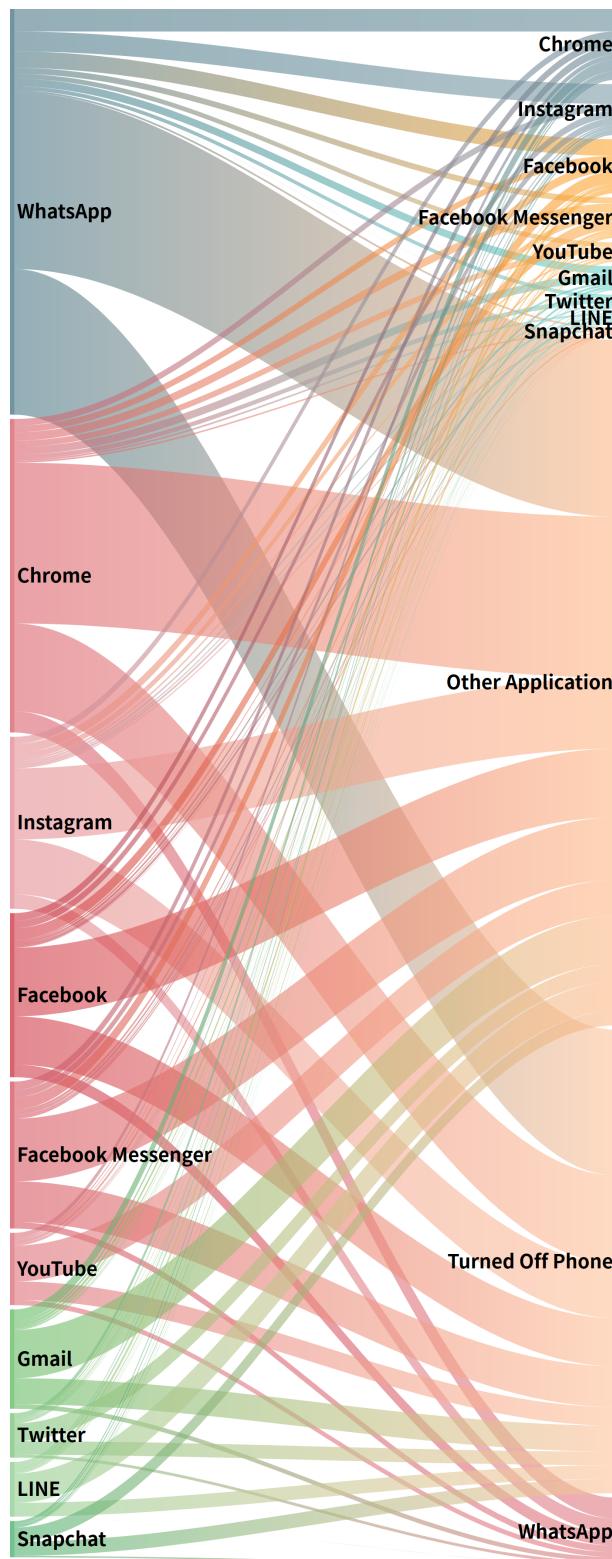


Figure 5.3: The top 10 goal apps with the most number of sessions on mobile are on the left. On the right is the distribution of where a user ends up immediately after.

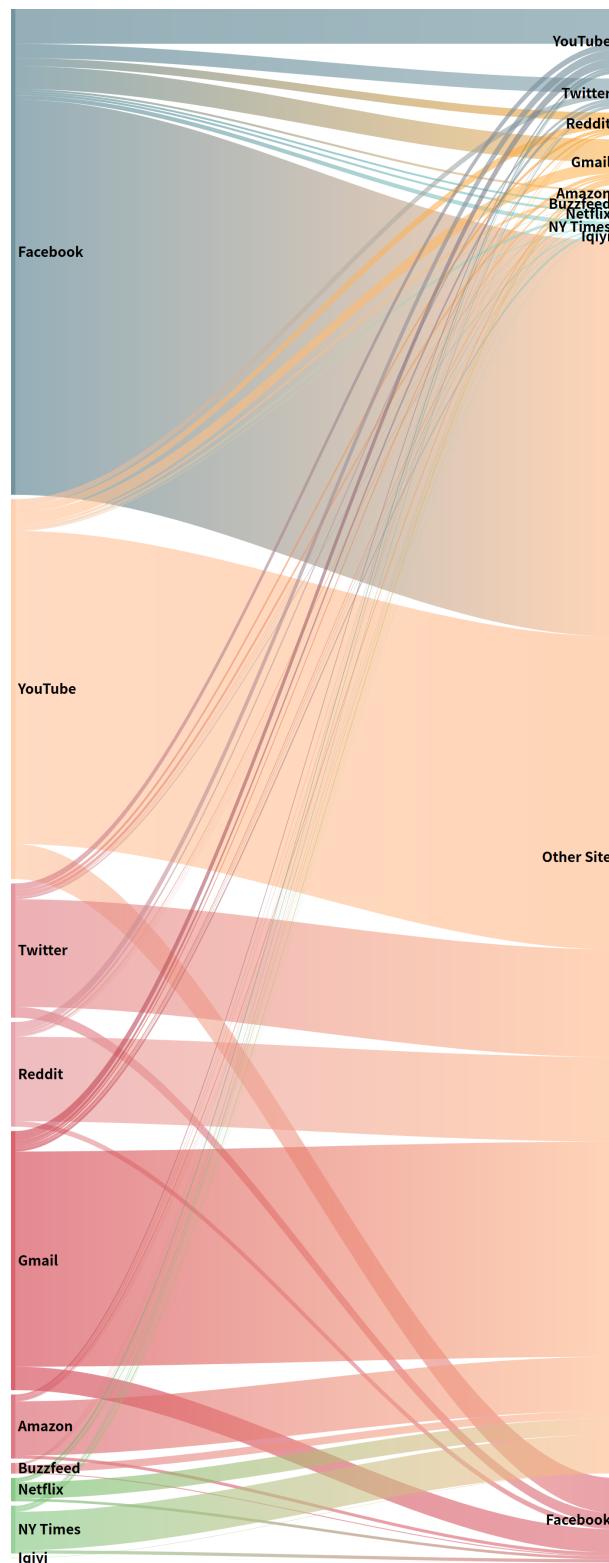


Figure 5.4: The top 10 goal apps with the most number of sessions on the browser are on the left. On the right is the distribution of where a user ends up immediately after.

## 5.8 Discussion

We found that productivity interventions on the browser also reduced time on sites other than the targeted sites, but there was no such effect on mobile or cross-device.

We believe the reason we observed reduction in time on non-goal sites on the browser is several of the most popular goal sites – such as Facebook, Reddit, Twitter – are filled with hyperlinks to other sites, and hence drive traffic to them. For example, if an intervention makes a user spend less on their Facebook feed, they are going to be less likely to stumble upon a New York Times article, hence the Facebook-reducing intervention may also reduce time on New York Times. Part of this may be a difference in how mobile applications work, compared to websites. Several mobile applications embed a web browser so that even if the user clicks a link, it will open within the same app. For example, Facebook is one such app, so if the user clicks on a New York Times link within the Facebook app, it is opened within the Facebook app’s built-in browser, so the time they spend reading that article will still be counted towards Facebook app usage.

One possible reason for differences between mobile and web is that the apps users choose to reduce time on in each two platform differ (e.g., messaging apps on Android vs. link aggregators on Chrome). There also exist differences in typical interaction styles (short, notification-driven sessions on Android [117], vs. longer sessions resulting from self-interruption on Chrome). 85% of the apps that Android users frequently chose to reduce time on are for messaging (WhatsApp, Instagram, Facebook Messenger, Twitter, LINE, Snapchat), where a characteristic interaction is receiving a message, unlocking the phone to read it and reply, then turning off the screen (as shown in Figure 5.3). Thus, users would not be drawn to other apps during this interaction. In contrast, with the Chrome version, the most selected sites are Facebook, YouTube, Reddit, and Twitter, 75% of which are aggregators of links to other sites. The number of daily sessions per app is also greater on Android, though sessions are longer on average on Chrome, and stopping using the browser after a session ends occurs less on Chrome. Thus, the browser-based interactions users were using HabitLab to reduce are not short messaging-driven spurts that end with turning off the screen as on mobile, but rather long sessions of surfing through link aggregators ending with going to another site. So, a proposed mechanism: interventions short-circuit browsing long browser-based sessions, but mobile sessions are already short.

This work brings about implications for designing interventions. Namely, we should consider not only the immediate interaction and its immediately measurable effects, but its longer-term effects in the context of the broader workflow. For example, consider 2 interventions for Facebook: 1) asks users to return to the home screen, vs 2) asks users to turn off the screen. Assuming similar rates of compliance, we would expect that measuring the effects on time spent on Facebook in isolation will show no difference between them. However, if we consider that going to the home screen can lead to users opening other apps, we might predict that a holistic measurement that includes effects on other apps as well will prefer 2) over 1). Or if designing interventions to reduce snacking, should we: a) ask participants to not eat anything until their next meal, or b) give them gum instead? While calorie intake from the immediate interaction would favor a), b) may prevent future snacking down the line. That said, in many cases, interventions are indeed

isolated in their effects, and can even have beneficial effects elsewhere.

The findings of this chapter about time redistribution have a positive tone – we did not observe negative side effects of productivity interventions, which would have been predicted if users were using their devices to replenish willpower when exhausted. Perhaps one speculative explanation is that in the context of device usage, diminished willpower results in the user opening a site or visiting an app, but actually spending time on the site or app does not replenish this willpower – only the initial act of opening the site or app does. This may have interesting implications if it is true in other domains as well – for instance, if the user is on a diet and has a craving for doughnuts, would an intervention preventing them from eating a doughnut also suppress cravings for other fattening foods as well? If they give in to the craving, would stopping them after the first bite leave their craving equally satisfied as if we let them eat the whole doughnut?

## 5.9 Conclusion

In this chapter we have compared three hypotheses for how productivity interventions influence time spent on sites, apps, and devices other than the ones they are targeting. Productivity interventions may have no effect on other goals (*isolated effects*), they may cause time to be redistributed to other unproductive goals (*redistribution*), or they may cause a reduction in time spent on other unproductive goals (*reduction*).

We adjudicated between these hypotheses by varying the frequency of productivity interventions on goals that users set in the HabitLab browser extension and mobile app. When interventions were more frequent, users spent less time on their goal sites and apps, showing that the productivity interventions were effective. We also defined a metric of intensity that captures frequency of interventions within device, and investigated the effects of varying intensity of interventions for other apps/sites, on time spent on an app/site. The result differed by device: on the browser we observed a global reduction effect, with time on non-goal sites decreasing with increasing intensity of interventions. However, on mobile we observed no effect. We believe these differences are caused by differing usage patterns and platform differences: websites drive traffic to other websites via hyperlinks, but mobile apps try to keep users remaining on the app.

We have shown that while productivity interventions can sometimes have effects on usage of other, non-targeted sites and apps, they are often isolated in their effects. Hence, when designing for behavior change, while we should be careful about our measurements and the possibility of unintended side effects, in the context of productivity interventions it appears that targeting individual productivity goals does not cause substantial negative second-order effects.

# **Chapter 6**

## **Discussion**

This thesis advances a vision of large-scale, in-the-wild experimentation which can simultaneously gather insights about behavior change across a large body of volunteer users, while at the same time providing users with a useful behavior change tool that they will voluntary use. These insights can then be used to create more effective behavior change system. This vision fits into a larger design space of in-the-wild research that attempts to find an intersection between the needs and incentives of end users, and the scientific needs of researchers. In this chapter will discuss our design principles for in-the-wild experimentation, our visions for the future of in-the-wild behavior change experimentation systems, as well as limitations of our general approach.

### **6.1 Design Principles for In-the-wild Experimentation**

Our design principles with HabitLab focused on maximizing user retention rates, which we believe helped us with growth. We did so by giving users control of their interventions, providing visually appealing and unobtrusive interventions, and avoiding intrusive surveys and experience sampling as much as possible.

We believe the key to our success with HabitLab was our approach of prioritizing user experience, retention, and growth. When conflicts between user experience and other research goals emerged, we would quantify the cost by conducting A/B tests to see the effects on retention, and opted for experiment designs which would give us the most useful data, with the least cost in retention. Our choice of studies to run, and research questions to pursue, was also influenced by this tradeoff – we chose research questions which would require minimal degradation in the user experience, and opted to not pursue those that would require extensive experience sampling and result in user dissatisfaction.

We believe there are many parallels between building in-the-wild experimentation systems, and monetizing commercial products, and researchers in this space can potentially learn much from the world of monetization. Specifically, from a user growth perspective in a commercial product, it is often best to prioritize user growth initially, and prioritize monetization later – and we see this play out in the business

strategies of many consumer-oriented startups. Likewise, when building HabitLab, we focused our early efforts on many features that were requested by users and helped retention, but were unrelated to any research questions – these early investments towards user growth assisted in ensuring we later had a sufficiently large active userbase to run our experiments. That said, with commercial products, the product and monetization strategy must be co-designed, so that the monetization strategy is aligned with user goals, and does not drive away users. Likewise, in the context of in-the-wild experimentation systems, research questions and the system must likewise be co-designed, so that the experiments do not drive away users.

## 6.2 Visions for the Future of Behavior Change Systems

What would an ideal behavior change system look like? Based on our studies with HabitLab, we believe it would be minimally intrusive, require minimal configuration, and yet be responsive to user preferences. It would include a variety of interventions, but the specifics of the interventions the user sees would depend highly on user preferences, contexts, and their goals.

A major advantage we had with the domain we chose – online behavior change – is that we could push our interventions automatically to users on every visit, without requiring any interaction on their part. This may be considerably more difficult to realize in other behavior change domains, where it may be more difficult to sense when the targeted activity is taking place. For instance, dieting apps unable to detect when the user is about to eat may require the user to explicitly open the app to indicate when and what they are eating, so different design strategies – such as unprompted push notifications that some might consider to be intrusive – may be needed for other behavior change domains.

That said, we envision that with the increasing ubiquity of sensors, augmented reality, and wearable devices, this objective of sensing behaviors in-the-wild and responding to them with interventions will become increasingly easy to realize. Once we are better able to sense the world, and can deploy interventions to the real world beyond our phones and browsers, a number of common behavior change goals – such as sleep, exercise, health, and dieting – become amenable to behavior change experimentation. We can envision, for example, using augmented reality to make our unhealthy food seem less appealing, and making our healthy food seem more appealing. Such an experiment can be deployed across a variety of users throughout the world, applying various filters to our food images as seen through our AR devices, and measuring the outcomes to determine the most effective filters. Another example might be determining what interventions are most effective to get a user to burn the most calories – some users may be motivated by social proof and competition from peers, others may be motivated by financial rewards, etc. We believe that compared to past behavior change studies, these new modalities of sensing outcomes and deploying interventions will open a the gates to much wider and potentially more effective interventions, as well as lowering the barrier to participation, enabling experimentation to be done at large scales.

### 6.3 Limitations

Given we conducted these studies in the context of reducing time spent online and on phones, results may not necessarily generalize to other behavior change domains. That said, with our increasing ability to sense our environment via sensors in our phones, smartwatches, and IoT devices, many of the paradigms we used in HabitLab can be applied to other domains as well. For instance, in the fitness domain, if we can sense users' physical activity levels via a smartwatch, we can experiment with various interventions that prompt users to exercise by playing audio messages or sending notifications. This hypothetical in-the-wild experimentation platform for fitness could potentially work analogously to HabitLab, running studies to find intervention strategies that work well to increase physical activity levels. The increasing ubiquity of sensors in the physical world make this paradigm of in-the-wild behavior change increasingly realistic and possible in domains outside online behavior change.

With regards to the general modality of in-the-wild behavior change experimentation with voluntary users, a major limitation is that we can only study behaviors that users truly want to change. For instance, if our behavior change objective is to cause users to gamble more, or spend more money shopping, users may likely be less willing to explicitly opt into voluntarily participating in such an experiment. Even if the behavior change objective is one that may appeal to users – for example, losing weight – certain types of interventions may be difficult to run accurate experiments with an in-the-wild, voluntary population. For example, if the intervention is to force the user to fast without any ability to opt-out, the user may end up interfering with sensors so they cannot detect them eating. Likewise, we suspect that if HabitLab were to start blocking sites entirely, users may be tempted to bypass the system by using another device or browser. Thus, we believe that when the incentives of the researchers and users do not align well, other means of experimentation – such as using compensated users in controlled lab settings, or forcing them to participate as the precondition for using a system – will still be more suitable.

# Chapter 7

## Conclusion

In this thesis we have proposed a paradigm of in-the-wild experimentation to gain insights about behavior change, and have created a platform, HabitLab, to realize this vision in the context of helping users reduce their time online and on their phones. We also conducted a set of studies on HabitLab which illustrate that we can make novel findings with the system.

The first set of studies we ran with HabitLab investigated whether interventions decline in effectiveness over time. We found that interventions decline in effectiveness if the same intervention is repeatedly shown, and that a strategy of rotating between different interventions can help improve the effectiveness. While this comes at the cost of increased attrition, most likely due to users having incorrect mental models, we can reduce this attrition via a simple design shown when a new intervention is introduced.

The second set of studies investigated whether interventions that help save time on one site, app, or device influence time spent elsewhere. We found that on the browser, reducing time on one site has a beneficial side effect of reducing time elsewhere. We believe this is due to reducing time on aggregator sites that drive traffic to other sites. On phones, however, we did not observe any side effects of reducing time on one app on other apps. We also did not observe any cross-device effects.

There is a large opportunity for behavior change research through big data and crowdsourcing that has been under-explored due to the paucity of large-scale deployments of research systems. Could we predict which interventions will work well for a new user, before they even start using the system? Could we automatically deploy and test modified versions of interventions, to hill-climb our way to more effective ones? Could we enlist an engaged user community to come up with, generate, and test new interventions for the long-tail of behavior change goals that designers had never even thought of? These can be realized with machine learning and crowdsourcing techniques, but there have not been appropriate communities for an in-the-wild deployment. We hope HabitLab will provide such a platform to realize this vision of community-driven behavior change research in the wild.

## **Appendix A**

# **Replication of Chapter 4, Study 1 Findings using Session Level Measurements**

This appendix replicates the findings of Study 1 using an alternative method of analysis, looking at time on site per session rather than day.

Time on site per session is measured as the total time the user was actively using a site in a browser tab, from when they visited the site until they closed the tab. If the user switches tabs to a different site, the time spent on the other site is not counted towards the current session time.

To determine whether the user is actively using a target site, we use Chrome's internal definition of active – the browser window and tab is focused, the computer screen is on, and there has been mouse or keyboard activity on the tab within the past minute. Because time data is not normally distributed, we adopt a common practice of log-transforming the time data prior to analysis.

### **A.1 Effectiveness of interventions over time**

The likelihood ratio test confirms that the number of times a user has seen an intervention affected the log of time spent on a domain per session ( $\chi^2(1) = 6.69, p < 0.01$ ), supporting H1. Each time the intervention has been previously seen increased the log time spent by 0.05633 (Table A.1). By exponentiating the log estimates, this translates into an increase of 5.8% on top of a baseline 46 seconds per session for each additional time the user saw the intervention during the study.

An alternative method of analysis, where we measure the raw number of times the intervention has been seen instead of the number of days it has been seen, yields the same results. Restricting analysis to just Facebook also yields the same results.

Table A.1: Within the static condition, interventions decline in effectiveness, with longer visit lengths with increasing larger number of days since it was first observed.

	<i>Dependent variable:</i>
	Log time spent per session
Number of days the intervention has been seen	0.056*** (0.021)
(Intercept)	3.826*** (0.143)
Observations	1,007

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

## **Appendix B**

# **List of Interventions used in Chapter 4 Studies**

The following is the list of interventions used for the studies in Chapter 4, showing the intervention name and description as seen by the end user. As this study was run before the mobile version of HabitLab existed, these interventions are all for the browser version.

There are 27 interventions total: 7 generic interventions that can be used on all sites, 5 interventions designed specifically for Facebook, and additional ones designed specifically for YouTube, Reddit, Twitter, Netflix, Gmail, Amazon, iQiyi, and Youku

Generic interventions that can be used on all sites:

- Minute Watch: Notifies you of time spent every minute
- Supervisor: Shows time spent on site at the top of screen
- Scroll Freezer: Freezes scrolling after a certain amount of scrolls
- Stat Whiz: Show time spent and visit count each visit
- GateKeeper: Makes you wait a few seconds before visiting
- 1Min Assassin: Closes tab after 60 seconds
- Bouncer: Asks how long you want to spend on site this visit

Facebook-specific interventions:

- Time Injector: Injects timer into the Facebook feed
- Feed Eater: Removes the Facebook news feed
- TimeKeeper: Notifies you of time spent in the corner of your desktop
- No Comment: Removes Facebook comments
- Clickbait Mosaic: Removes clickbait from the news feed

Youtube-specific interventions:

- Sidekicker: Remove sidebar links

- Think Twice: Prompt the user before watching a video
- No Comment: Removes comment section

Netflix-specific interventions:

- Fun Facts: Gives you a fact and links an article on the effect of TV
- Alarm Clock: Asks the user to set an alarm before watching a show
- Stop Autoplay: Stops the site from automatically playing the next video

Reddit-specific interventions:

- Comment Remover: Removes Reddit comments
- Mission Objective: Asks what you aim to do this visit and puts a reminder up

Youku-specific interventions

- Think Twice: Prompt the user before watching a video
- Sidekicker: Remove sidebar links

iQiyi-specific interventions

- Think Twice: Prompt the user before watching a video
- Sidekicker: Remove sidebar links

Twitter-specific interventions:

- Feed Eater: Removes the Twitter news feed

Amazon-specific interventions:

- No Recs: Hides recommendations

Gmail-specific interventions

- Speedbump: Delays the arrival of new emails

## **Appendix C**

# **List of Interventions used in Chapter 5 Studies**

The following is the list of interventions used for the studies in Chapter 5, showing the intervention name and description as seen by the end user.

### **C.1 List of Browser Interventions**

Generic interventions that can be used on all sites:

- Minute Watch: Notifies you of time spent every minute
- Supervisor: Shows time spent on site at the top of screen
- Scroll Freezer: Freezes scrolling after a certain amount of scrolls
- Stat Whiz: Show time spent and visit count each visit
- GateKeeper: Makes you wait a few seconds before visiting
- 1Min Assassin: Closes tab after 60 seconds
- Bouncer: Asks how long you want to spend on site this visit

Facebook-specific interventions:

- Time Injector: Injects timer into the Facebook feed
- Feed Eater: Removes the Facebook news feed
- TimeKeeper: Notifies you of time spent in the corner of your desktop
- No Comment: Removes Facebook comments
- Clickbait Mosaic: Removes clickbait from the news feed

Youtube-specific interventions:

- Sidekicker: Remove sidebar links

- Think Twice: Prompt the user before watching a video
- No Comment: Removes comment section

Netflix-specific interventions:

- Fun Facts: Gives you a fact and links an article on the effect of TV
- Alarm Clock: Asks the user to set an alarm before watching a show
- Stop Autoplay: Stops the site from automatically playing the next video

Reddit-specific interventions:

- Comment Remover: Removes Reddit comments
- Mission Objective: Asks what you aim to do this visit and puts a reminder up

Youku-specific interventions

- Think Twice: Prompt the user before watching a video
- Sidekicker: Remove sidebar links

iQiyi-specific interventions

- Think Twice: Prompt the user before watching a video
- Sidekicker: Remove sidebar links

Twitter-specific interventions:

- Feed Eater: Removes the Twitter news feed

Amazon-specific interventions:

- No Recs: Hides recommendations

Gmail-specific interventions

- Speedbump: Delays the arrival of new emails

## C.2 List of Mobile Interventions

All mobile interventions are generic, that is they can be used on any app.

- At it Again: Sends a pop up with your app visit count.
- Progress Report: Sends a pop up with today's total usage for a certain app
- Red Alert!: Sends a notification with today's total usage for a certain app
- Repeat Offender: Sends a notification with your app visit count
- All in All: Pops a dialog with the day's total time on the current app
- Back To Target: Suggests you to visit a target app
- Counting on You: Puts a timer on screen in watchlisted apps
- Man Overboard! Shows a dialog with your app visit count
- No Peeking!: Asks for confirmation before opening watchlisted apps
- Wait Up! Pause for 10 seconds before entering an app
- Your Better Half: Sends a pop up to go to a target app
- Look on the Bright Side: Dim the screen a little at a time
- Take Your Pick: Select how long you want to spend on an app
- The Final Countdown: On screen timer that closes the app when time runs out

The following interventions apply across the device as a whole, not individual applications.

- How Time Flies!: Sends a pop up message with current app visit length
- Knock Knock: Sends a pop up with your glance count for the day
- Long Time No See: Sends pop up with your phone usage for the day
- Call it a Day: Sends notification with phone usage for the day
- Easy on the Eyes: Sends notification with glance count for the day
- Hello, Old Friend: Sends notification with unlock count for the day
- The Clock is Ticking: Sends a notification with the current app visit duration
- En Garde: Pops a dialog with the day's total unlock count
- Hold the Phone: Show dialog with phone usage for the day
- Long Story Short: Pops a dialog with the visit time for the current app
- Quote reminder: Show quote upon opening app
- Time Reminder: Show dialog with phone usage for the day
- Take Your Pick: Select how long you want to spend on an app

## Appendix D

# User-Contributed Intervention Ideas

Here is a complete list of intervention ideas that users have submitted via HabitLab:

- 1: Buzzfeed : n ,
- 2: Generic: "In this [x] minutes on [site name] you would have... [fill in the gaps]
- 3: Generic: Disable news feed on LinkedIn, my biggest time waster
- 4: Amazon: Given max, if user spent money>max remind him that is useless to go to Amazon!
- 5: Facebook: only allow access to facebook messenger page + only allow access to Events page
- 6: Reddit: Just break the computer
- 7: Generic: motivational quotes
- 8: Generic: confirmation before loading the page, with a certain time forcing you to think
- 9: Facebook: Remind us Facebook admitted that it has always been a data mining Co. Period ...

- 10: Youtube: Nudge after watching X minutes of video
- 11: Generic: before opening the site , display a (personal) list of alternative things to do
- 12: Generic: mom simulator: timed popups , are you using site because <x>? do <y> instead
- 13: Twitter: remove cats , categorized very popular BS feeds such as Only in Russia or similar
- 14: Generic: Automatic redirects to a more useful (user chosen?) site
- 15: Generic: Slow your scroll speed the longer you are on the site .
- 16: Netflix: Set time for length on website , after that time length , close tab .
- 17: Generic: Monitorize time spent on internet at all
- 18: Generic: Count limit number of posts , shares , reactions , etc . basically 'acts '
- 19: Generic: Ask when site opened if its related to goal.Limit tabs # , Use pomodore. timebox
- 20: Generic: Delay any tempting idea to search ask does it really make a difference in ur lif
- 21: Generic: <http://humanetech.com/>
- 22: Generic: I can suggest much more if if you enable me to . I have lots of ideas
- 23: Amazon: A budget reminder , so people don't end up overspending browsing random items .

- 24: Generic: stops you visiting a selected site for a while after your first visit
- 25: Youtube: Every minute it would display the amount of data used by youtube.
- 26: Generic: As I prep for tests , I'd love to get encouragement to spend time on some sites .
- 27: Amazon: It will tell you how much you spent and if you spend over \$50 , it block Amazon
- 28: Generic: Nudges applied to all bookmarks. e.g. to justify my visits to favourite websites
- 29: Generic: Generate a social media post announcing how much time you've wasted and where.
- 30: Generic: an hourglass of the mins I have to live , and how many of them are spent online
- 31: Generic: Prevent me from browsing sites during certain times.
- 32: Generic: Disable or limit clicks on outgoing links on a page
- 33: Amazon: Donate money to charity on your behalf
- 34: Generic: I just need a tracker for a total time spent in the internet .
- 35: Generic: Could you please design a tracker for a tomal time spent in the internet ?
- 36: Generic: I would like to b abl to change the size of the "timer ". Current one is too smal

- 37: Nytimes: Limit to two (or N) links clicked from the homepage each visit (or per hour)
- 38: Nytimes: Warning before loading the site for the second time in less than an hour
- 39: Generic: Remind me I just opened/closed the same damn tab 7m ago. For the 14th time today
- 40: Youtube: play asmr videos only
- 41: Generic: Add a blurry overlay to the screen that gets worse as time goes on.
- 42: Generic: For a porn website , add popup memes during the videos .
- 43: Facebook: Close the tab after a certain amount of time
- 44: Generic: Goals reminder: reminds you of your goals to promote better decisions
- 45: Netflix: Prompt user to confirm if they wish to continue watching before next auto play
- 46: Youtube: Force video to take entire window (not fullscreen , just larger theatre mode)
- 47: Youtube: Force disable autoplay
- 48: Youtube: Restrict the number of links that can be followed in one session (Wikipedia too )
- 49: Generic: 1 thing that u need to do & 1 thing you wish you could do if u had more time
- 50: Gmail: Limit the number of refreshes to inbox before auto-logging out

- 51: Generic: A popup reminder of why this is important to you, e.g. "your novel is waiting"
- 52: Generic: How about tracking time spent in Incognito mode?
- 53: Gmail: Self-set timer - e.g. reminder every 30 mins
- 54: Youtube: Reminder triggered by start of new video or pausing video
- 55: Facebook: Force you to wait at least 60 seconds before a finished comment is posted. PLZ.
- 56: Generic: Ask what your goal is for the visit
- 57: Generic: Ask how much time you want to spend on a site before the tab closes
- 58: Twitter: Make notification updates non-live - e.g. once per hour / 2 hours
- 59: Buzzfeed: redirect to home screen or site of choice
- 60: Facebook: redirect to random wiki
- 61: Netflix: close page at end of episode or when paused
- 62: Gmail: apply to-do list with task-based timer
- 63: Generic: Prompt for hourly pay rate. Express time wasted in dollars wasted.
- 64: Generic: display total weekly stats at top of page
- 65: Youtube: closes when the viewer watches a certain amount of videos

- 66: Amazon: Do you need what you are about to go here and look for? (any eCommerce, really)
- 67: Generic: Restricts how many times you can open a site. eg 10–20x (For me, a stocks site)
- 68: Generic: 1 minute assassin , but w/o stupid "add time" button . Clicking that feels amazing
- 69: Twitter: limit thread or hashtag to 20 tweets
- 70: Custom: limit comments to 1 page , or remove next page button
- 71: Facebook: Meditate: count n in breaths and n out breaths (e.g. n = 3–5) before entering
- 72: Youtube: Prevent constantly changing videos by showing a pop up
- 73: Generic: sync with Google Tasks (or whatever task manager) , show what is due next
- 74: Amazon: A counter that shows the number of clicks...the number of items you've viewed.
- 75: Generic: Sentences like "In the time you spend here weekly you could learn X in Y weeks"
- 76: Youtube: Lock the video interface and then slowly turn down the volume
- 77: Generic: Spanish please
- 78: Facebook: Only show 3 posts from the newsfeed , no more.
- 79: Twitter: Only show 10 tweets and don't load any additional ones when you scroll .

- 80: Generic: Changing end-of-day from 12am to 4am would be better for me , still up at 12am.
- 81: Reddit: Ask "why did you visit Reddit"? (A) Share something (B) Break. (C) etc ...
- 82: Generic: Scroll Reseter: Automatically scroll back to top of page after N scrolls .
- 83: Generic: Show link to some alternative website you could be spending your time on.
- 84: Generic: Disables clicking for the first 15 seconds on the website .
- 85: Generic: Delays 15 seconds between clicking a link and having that link open .
- 86: Facebook: Popup msg spend time with family and friends with calc time wasted(avg life - t)
- 87: Generic: Solve basic math problem
- 88: Generic: x min assassin aggregated per some webs (x set up by user)
- 89: Generic: A big text saying: You have some important things to do!
- 90: Generic: a large message appears blocking the screen content
- 91: Netflix: Ask you how many episode(s) you will be watching
- 92: Generic: Ask me: Is this educational?
- 93: Facebook: Remove News Feed for a Day
- 94: Generic: send a warning/stop after opening x number of links
- 95: Generic: nothing

96: Amazon: Ask the person if they NEED the product , or just WANT the product impulsively

97: Amazon:

98: Reddit: Selected subreddit – Only allow access to selected subreddits for a little time

99: Youtube: Prepare for IIT JEE , see only relaxation music and iit related lecturesvideos

100: Generic: Ask how much time I want to spend on the site , close the tab after time runs out

101: Twitter: Scroll freezer

102: Generic: An exploding burst of colour & glitter with text – time to go outside & play

103: Generic: An exploding burst of colour with the text 'time to go outside & play'

104: Amazon: Show a timer in the corner that says how much time you've spent browsing

105: Amazon: Count the number of products/pages you've clicked on since visiting the site

106: Generic: What is your spending limit for this website/purchase?

107: Generic: In general: offer sign in option first for extension , handy if more than 1 pc

108: Generic: Also , increase text limit nudge suggestion :P

109: Amazon: What you want to buy? And how much do you want to spend?  
And inject amount cart

110: Facebook: Blocks specific Facebook functions such as Feed and spamming notifications.

111: Youtube: stops you from being able to watch a video after watching a set amount of videos

112: Generic: Remove instagram feed

113: Facebook: Set time that you want the browser to block your Facebook

114: Generic: A video montage of people accomplishing things to show what you're missing

115: Generic: The screen turns blank every 10 seconds with a message

116: Generic: an updated list of things you could have accomplished in the time you've spent

117: Gmail: Limit or alert after 30 seconds on a "compose" screen.  
Emails should be short.

118: Generic: flip the screen upside down

119: Generic: Shake the screen (window) visually or temporarily black it out

120: Generic: A animated figure pops on the screen point to there watch and taps it.

121: Calm: Do something productive

122: Gmail: Read or Go for a walk

123: Youtube: Call a Friend

124: Generic: Not a nudge , but a timer so that this only works during work hours would be fab !

125: Generic: Combine with a pomodoro timer to make this a super productivity tool

126: Generic: Holy grail> an active hours function+pomodoro , that live syncs across devices

127: Twitter: hide retweets that are trending too quickly , as they 're likely time-wasters

128: Amazon: Ask what I am specifically shopping for

129: Generic: darken screen; large timer in center. press "space" to leave; ESC to stay

130: Generic: Please make this for Firefox

131: Twitter: Put fake tweets into feed asking with increasing frequency , "Why are you on Tw?"

132: Youtube: videos play faster , or ads play slower , gets progressively worse

133: Generic: freeze keyboard controls until the tab is removed after a set period of time

134: Facebook: Ask you a group of peoples feed to show only (eg: just family )

135: Generic: 5 minute assassin option(similar to the 1-minute assassin nudge)

136: Generic: multi nudges at the same time

137: Generic: 5 minute assassin timer nudge(similar to the 1 minute assassin nudge)& show time

138: Generic: When I am getting off track from the original reason I went online. Surf time

139: Generic: Flip the screen vertical

140: Amazon: Logs you out after a set amount of time.

141: Generic: ask how scrolls I need to do to block it

142: Gmail: sticky note pop out on your screen and say girl or boy you have 5 mins left

143: Twitter: Pop up that says , "Haven't you already read this?" (News services , too!)

144: Facebook: Simple questions: Did you find what you were looking for? Still here? etc .

145: Generic: Plays an alarm sound every x minutes , for example every 10 minutes you hear it .

146: Youtube: show time spent on youtube

147: Youtube: que no pongan anuncios en pleno video(en medio) del video

148: Youtube: stop auto-play of next video – replace with a prompt/timer /reminder if possible

149: Amazon: ask what online to buy , how much time needed , then prompt when approaching time

150: Generic: Telegram

151: Youtube: Close after 1 hours

152: Youtube: Only show the videos you need

153: Generic: Timer but flashes every X seconds in diff parts of screen

154: Generic: if you visited a website for the 20iest time in the last 3 hours , block it

155: Generic: blank screen intermittent after said time

156: Generic: feedeater& thinktwice (bcz user get there with purpose , but feed wastes) quora ,

157: Youtube: Block the tab at certain times that you can set up

158: Youtube: Is this for study or not ?

159: Youtube: Auto-pause a video after a certain amount of time and show your stats .

160: Generic: Pop ups every 5 minutes that remind you and ask if you should still be on there .

161: Generic: pictures to a meme or to a set picture which motivates people ,like family pictur

162: Youtube: Receive a warning after spending some specified amount of time

163: Generic: Every min, flash screen off for 5 seconds and remind them their goals + accompl .

164: Youtube: At a certain time , set by the user , the video will start stutter .

165: Youtube: Classify the video title (fun/useful) notifies user if he/she spends too much time

166: Generic: Make the remind/aim banner dark instead of white

167: Generic: Ability to set specific nudges on or off , I just like certain ones more.

168: Generic: Allow control of consumption by amount of data spent

169: Generic: For all websites , ask how long you want to be on the internet for all together

170: Youtube: Detect "Recommended" browsing and how many videos you've hoped on a row

171: Youtube: Click (twice or let users wait) to show the video recommendations on the side

172: Facebook: Ask math question before opening site .

173: Generic: Linkedin

174: Youtube: Get rid of sidebar

175: Twitter: For a certain time frame , have only a certain subset of tweets appear on feed .

176: Twitter: Tick down the # of links , replies , etc . you click on then freeze for some time .

177: Facebook: Hide repeated posts or hide posts that has been already displayed

178: Generic: Ask to stop or close tab

- 179: Generic: Higher Self. Display a prompt encouraging user to reflect on values
- 180: Generic: Link Lock: Sets (x) amount of links/videos user can click per day per site
- 181: Generic: The time wizard , but non-intrusive . The Mission Objective , but on specific sites
- 182: Youtube: Vomiting audio
- 183: Generic: Vote if tab "created idea" or "killed time" on close; report ideas by tab title
- 184: Facebook: I want the few second delay just for facebook
- 185: Generic: A notification to bring user back to the browser and stop surfing offline files
- 186: Generic: Go to bed (with insomnia option)
- 187: Amazon: Asks you what you are here to purchase
- 188: Reddit: Shows only the top 10 threads in the 'popular' section
- 189: Facebook: hide news feed
- 190: Generic: Personalized messages on screen (ex. "Don't you want to go to medical school??")
- 191: Generic: Change all images to Nicolas Cage face (Like chrome's ncage extension)
- 192: Generic: every minute , the webpage becomes gradually more pixelated
- 193: Generic: Put links into a TODO -like list instead of opening them. This helps focusing .

194: Generic: After a configurable amount of time , close website and don 't let me reopen it .

195: Generic: Lock to this page  
Helps focusing on learning material. Block other pages.

196: Youtube: Limited amount of videos one can watc , as sometimes you need to watch for work .

197: Youtube: Block certain channels

198: Reddit: Remove the "next page" link on the old reddit , or block auto-scrolling on new

199: Reddit: Ask how many minutes to keep the tab open before killing it

200: Generic: pixels start getting blacked out

201: Reddit: Show only the first page of each subreddit and hide (or show less )

202: Generic: Time limit for a group of websites eg 10mins all social media , then no access

203: Buzzfeed: for them to stop being feminists

204: Ted: stop

205: Duolingo: learn spanish or die

206: Facebook: hello fellow white mothers

207: Facebook: HabitLab could post amount of time spent and visit count on FB wall

- 208: Generic: Progress bar that moves forward as user is productive and backward when not
- 209: Youtube: Turn off the sound for youtube/Pause the video , or put the video in slow motion
- 210: Youtube: Pauses the video and reminds you of a previously set reminder .
- 211: Youtube: Make a list for how much time you want to spend on each site and be reminded
- 212: Netflix: Shows you how many hours you have wasted on that particular show
- 213: Netflix: Close tab after preselected numer of episodes watched
- 214: Netflix: Lock episodes for series to one per day
- 215: Generic: Break time – set your timer for your break between tasks
- 216: Amazon: Shows your monthly/weekly/yearly purchase history dollar amount \$\$\$.
- 217: Amazon: Removes Buy buttons , replaces them with Add to List so you have to wait & think
- 218: Youtube: Ask to minimize tab , so user can just listen not watch ( for music or talks )
- 219: Buzzfeed: Tells you that you have better things to do and after that close after 60 second
- 220: Generic: Ask how are you feeling right now? or Is there a task you are avoiding?
- 221: Generic: Could you use a break – get up and take a walk .

- 222: Generic: Turns the entire screen white gives you a life quote and asks to continue
- 223: Twitter: Show only your notifications; clicking on any links forces closes Twitter.
- 224: Generic: Switch to 1st tab in browser window after specified time.
- 225: Generic: Mindfulness popup, after specific time elapsed.
- 226: Generic: After a certain amount of time, asks what your purpose is for still being there
- 227: Generic: make it black and white
- 228: Twitter: Don't allow feed refreshes
- 229: Twitter: Set a max time allowed per day, week, month
- 230: Youtube: When a video finishes Can you remove suggestions or next video?Or even playlists
- 231: Generic: A pomodoro technique complement. Work 20' and get your 5' treat.
- 232: Youtube: Close after a certain amount of time and restrict access until you say
- 233: Youtube: mode music please
- 234: Youtube: mode music please
- 235: Gmail: (if enabled) remove the "new messages" icon from social and promotions tab
- 236: Duolingo: Mejor es aprender otro idioma

237: Duolingo: aprender

238: Youtube: Ask me before what type of videos I want to watch/block and work over this list

239: Generic: PowerUSE: Utility-Span-Expiry. State purpose, make estimate, agree hardstop end.

240: Youtube: Music only mode. Disable videos, but allow music. To listen while working.

241: Facebook: Include Facebook Messenger ([www.messenger.com](http://www.messenger.com)) blocking messages or a time limit

242: Generic: advertising videos goggle up my RAM, slowing my page loading. Stop them please.

243: Generic: Top-level comments only. Removes all reply comments.

244: Generic: Block access to the website for a period of time

245: Netflix: Remove Suggested Shows/Movies

246: Twitter: Remove Likes/RT for your Tweets and removes comments for other people's Tweets

247: Youtube: Hide the comment section. A total time-killer.

248: Generic: Mood Tracker

249: Generic: Describe mood before clicking on a website

250: Generic: If on a website for too long, play a recording of yourself saying other ideas

251: Generic: A reminder that the work won't disappear , but these other distractions can wait.

252: Youtube: Only show the top 5 comments (and 3 replies per comment).

## Appendix E

# User Feedback on GitHub Issues

Here is a complete list of feedback that users left us via the feedback interfaces in HabitLab, where users agreed to have them be publicly shared on GitHub Issues. We removed all examples with any identifying information. Many of these are associated with images and screenshots, which can be viewed by clicking the GitHub link.

<https://github.com/habitlab/habitlab/issues/641>

Hi – I love your app, but I have a few suggestions / requests .

I think it would be great if there was the option for greater control to select multiple nudges to function every time. In my scenario , the website I want to control is Youtube. I use Youtube a lot whilst I'm working and being productive for various tutorials , downloading copyright fee assets and so on. However, the recommended videos often push click bait "trending" content at me, and having gone on Youtube to find a tutorial on solving a certain problem , you suddenly find your self 5 minutes into a "You won't BELIEVE what Gordan Ramsey says to this Chef" or similar rubbish. I'd really like to be able to use the Feed Diet , Sidekicker , Supervisor , and No Comment at the same time . I feel like your app has everything I need , but I can't use it all at the same time :)

Best ,

Ben

<https://github.com/habitlab/habitlab/issues/639>

The disabling of autoplay makes me skip the recap of Jane The Virgin . However, that series includes new information and new jokes in every recap , so I always watch each recap even if I 've just seen the previous episode . With autoplay on I can 't watch the recap even if I 've just logged in to Netflix , and even if I 'm "rewinding" back to the beginning of the episode each time . It skips it again .

<https://github.com/habitlab/habitlab/issues/638>

it keeps asking me "how much do you want me to bother you??" and i am tired of answering this question . very cool extension that i used for like a year but something seems to have gone wrong so now i 'm uninstalling :(

<https://github.com/habitlab/habitlab/issues/637>

I really wish there was a setting where I could use a nudge on every website except what 's on a whitelist , because I always find a new place to waste time .

<https://github.com/habitlab/habitlab/issues/635>

Is there a way to set a default level of "how aggressive" for sites we want to reduce time on instead of choosing each time we load the site ? ie , if I 'm in the process of pulling up facebook , I 'm much less likely to choose "heavy-handed" etc in the moment than I might in a more clear-headed moment ahead of time

<https://github.com/habitlab/habitlab/issues/634>

Disabling HabitLab due to excessive CPU usage . Right now the Chrome browser 's TaskManage , CPU usage column shows HabitLab using 10–15% CPU . Ouch ! Adios !

<https://github.com/habitlab/habitlab/issues/633>

What about separating out Amazon from it's Kindle page (`read.amazon.com`) and it music page (`music.amazon.com`) because I enjoy listening to music when I type, and (sometimes) I read books on my Kindle from the website for work.

<https://github.com/habitlab/habitlab/issues/632>

You should be able to group sites and provide overall limits and timers across the category. For instance, Netflix and YouTube would be considered 'media streaming' allowing the user to set a goal of say an hour of streaming a day. Otherwise, I can personally see me spreading my viewing over multiple different websites to (pretend to ) keep within the goals .

<https://github.com/habitlab/habitlab/issues/631>

It's annoying that it asks always what nudge I want, when you open a website. Please stop this!

<https://github.com/habitlab/habitlab/issues/629>

It's annoying that it asks always what nudge I want, when you open a website. Please stop this!

<https://github.com/habitlab/habitlab/issues/628>

I have told this app again and again that I want to turn off 2 nudges , freeze scrolling , and setting the number of minutes I want to spend on a website in advance. HOWEVER IT KEEPS COMING BACK and I am almost about to delete this despite loving ever other aspect of it . FIX THAT ASAP. OR don't say you can remove it if you can't .

<https://github.com/habitlab/habitlab/issues/627>

Hey! Would be nice to have displayed the total amount of time spent on the web. Great work, people! Thank you for making this :)

<https://github.com/habitlab/habitlab/issues/624>

When I click sign in to sync nothing happens. I closed and opened new tabs and such and its still not working so I can't sign in

<https://github.com/habitlab/habitlab/issues/620>

Youtube nudges not working as expected. Sidebar and comments turned on and showing normally. How many nudges can I active simultaneously?

<https://github.com/habitlab/habitlab/issues/618>

I want to see my history and detailed results for longer durations , such as weeks or months

<https://github.com/habitlab/habitlab/issues/617>

I came to Facebook to check notifications for events , but the scroll freezer hides the entire top bar. And the search bar too! I can't manage the events that I came here to manage. The other HabitLab stuff is good though!

<https://github.com/habitlab/habitlab/issues/616>

Love HabitLab so far! However, I am unable to log in on the extension . I press the button and it does nothing .

<https://github.com/habitlab/habitlab/issues/615>

Thank you so much. I downloaded this to track my work time ( work on social media) and stumbled across the feed diet during one of the random cycles on youtube. I have never felt my mind clear like that just by not having the feed!! amazing extension , the possibilities are

<https://github.com/habitlab/habitlab/issues/613>

Nudges should NOT cover the page , they should possibly push the whole page down by the amount of space needed by the nudge:

most apps , like twitter and facebook , have their buttons at the top , and your nudges simply cover those and make the site unusable , so maybe one wants to do a quick action on the sites bar and leave , but the nudge bar is in the way , so one is forced to close the nudge to access the function of the site and get on with it . But then the nudge is closed

Another possible solution is give the option to hide the nudge for 5–10 seconds instead of 1hour, but I prefer to see how the behaviour changes if the site usability is unencumbered

Mission Goal Nudges should allow at least to see what the page you were opening was. The objective can't be defined if you have multiple tabs open and you can't interpret from the url what the content you wanted to open was. Ideally it will be as blocking as it is now... so that it is the only action you can do, but you should be able to see the background page through for instance a darker screen, or a "peek" button that hides the nudge on rollover and shows it again as soon as you exit the button.

<https://github.com/habitlab/habitlab/issues/610>

<https://github.com/habitlab/habitlab/issues/607>

Show total browsing time including those pages not in top 5.

<https://github.com/habitlab/habitlab/issues/606>

Password for settings page, and nudge turn off buttons.

<https://github.com/habitlab/habitlab/issues/605>

Option to hide Turn off HabitLab button.

<https://github.com/habitlab/habitlab/issues/604>

Please add an option in settings to hide nudge turn off buttons (so they can only be turned off from settings page).

<https://github.com/habitlab/habitlab/issues/601>

On sites like news.google.com and reddit.com, you can click on links that take you to long articles on other domains where you can spend a significant chunk of time. Habit Lab doesn't track such changes now but really should!

<https://github.com/habitlab/habitlab/issues/597>

I want to see results per week and month – and to be able to compare each week / month. Also, an option to choose which day the week starts

<https://github.com/habitlab/habitlab/issues/594>

My nudges are not working most of time.

<https://github.com/habitlab/habitlab/issues/593>

i would like to suggest maybe you can make another tracking nudge that only allows you to stay on a certain website and cant open any other tabs

<https://github.com/habitlab/habitlab/issues/590>

Hi there ,

Sometimes , habit lab seems to turn off and I have no idea why. I haven't seen any nudges on facebook in the past week , but I didn't change any of my settings . This is the second time I've had this issue ( last time I had to reinstall habit lab to fix it) . What's the best way to fix this ?

<https://github.com/habitlab/habitlab/issues/589>

Add minutes , not percentages , so i can see how many minutes i wasted on certain sites !

<https://github.com/habitlab/habitlab/issues/587>

Banner says a different website than the one I'm actually on .

<https://github.com/habitlab/habitlab/issues/585>

It won't show nudges

<https://github.com/habitlab/habitlab/issues/583>

In general , I think the random is a good idea , but I quickly realize some of the random ones are ineffective for me. Rather than the random , it would be great if I could designate 1 feature for a particular website. For example , I know that facebook is a big time waster to me and moreso than others. I would love it if that was 1 one minute kick-off no matter what. Others (eg - Twitter) are less of a distraction for me, so the random wouldn't be as painful.

<https://github.com/habitlab/habitlab/issues/577>

For the past few days , it only seems to track YouTube for some reason .  
Is there a way I could fix this ?

<https://github.com/habitlab/habitlab/issues/576>

Hello ! I am not sure if I am understanding properly , but with the Bouncer nudge (my favourite) , once you have done it once in a day it never triggers again. It would be good for it to ask every time I go to a site , how long I want to spend on it. And if I exit the site , it refreshes and starts again .

<https://github.com/habitlab/habitlab/issues/572>

Nudges will frequently not show up when I visit facebook. Why is this ? I haven't accidentally turned them off .

<https://github.com/habitlab/habitlab/issues/571>

Banner is too large

<https://github.com/habitlab/habitlab/issues/570>

the nudges arent really working .... i have visited facebook multiple times now, but not have been nudged even once. I have enabled all the nudges as to see which one helps me the best , but its not nudging at all .

<https://github.com/habitlab/habitlab/issues/569>

None of the nudges are showing up. I'm not sure how to track down the issue. I've pasted the contents of the javascript console from a visit to YouTube here <https://pastebin.com/ZyPYB2Tw> in case it helps track down the issue. Does HabitLab not work alongside adblockers or ghostery or something like that?

<https://github.com/habitlab/habitlab/issues/568>

Thanks

<https://github.com/habitlab/habitlab/issues/567>

your "how should we handle this" modal thing disappears on most pages before any human brain can be expected to click the options

<https://github.com/habitlab/habitlab/issues/566>

The extension doesn't track subdomains of a main domain properly.

Suppose I add xyz.com as a filter and then it directs me to abc.xyz.com, the filter doesn't provide the needed nudges. This can be helpful if implemented.

<https://github.com/habitlab/habitlab/issues/565>

HabitLab extension in Chrome is using a lot of CPU

<https://github.com/habitlab/habitlab/issues/564>

Please add a login option at the start of the onboarding process to pull down any sync'ed data. I use a number of browsing logins on a number of computers and don't want to have to choose the list of sites every time. Also, a 'logout' option would be good. Also also, an email+password login would be handy for those who don't want to use a google account for this (or who want to share tracking and prefs across work and personal browser sessions)

<https://github.com/habitlab/habitlab/issues/563>

I believe settings across similar devices (laptop and desktop versus tablet and smartphone) should sync.

<https://github.com/habitlab/habitlab/issues/562>

I would love it if it would give me a running tally of my total minutes spent in addition to the "Today's five most visited sites by minutes spent" Thanks and cheers!

<https://github.com/habitlab/habitlab/issues/561>

How much time did I spend on what tabs, last week? Where are the totals?

<https://github.com/habitlab/habitlab/issues/560>

Think twice nudge for youtube does not work.

<https://github.com/habitlab/habitlab/issues/558>

How do I sign in to sync after the onboarding process? (Sign in to sync didn't work when I onboarded)

<https://github.com/habitlab/habitlab/issues/557>

Why do I not always receive a nudge when I visit facebook, I keep compulsively checking, and hoping a nudge will remind me, but I don't seem to be seeing any.

<https://github.com/habitlab/habitlab/issues/554>

I used two computers. On both computers, I am logged into Chrome with the same account as well as logged into HabitLab. However, my settings are not in sync.

<https://github.com/habitlab/habitlab/issues/553>

Can you make a whitelist version?

<https://github.com/habitlab/habitlab/issues/550>

how about running it in incognito mode

<https://github.com/habitlab/habitlab/issues/548>

Please add a whitelist for stricter management, so we add the sites we DO want to access.

<https://github.com/habitlab/habitlab/issues/547>

Feed Eater Bug– if I have the feed eater feature enabled, every time I open facebook, there will be an alert window along the lines of "How aggressive would you like HabitLab to be in helping you reduce your time spent this visit?", which gets tiring when you have to open facebook a lot of times for personal matters (not time wasting stuff I swear)

Update: Actually ignore what I just said about the Feed Eater bug, even with the feature off the issue continues on.

<https://github.com/habitlab/habitlab/issues/546>

Feed Eater Bug– if I have the feed eater feature enabled, every time I open facebook, there will be an alert window along the lines of "How aggressive would you like HabitLab to be in helping you reduce your time spent this visit?", which gets tiring when you have to open facebook a lot of times for personal matters (not time wasting stuff I swear)

<https://github.com/habitlab/habitlab/issues/545>

Feed Eater Bug

<https://github.com/habitlab/habitlab/issues/544>

Please let me have the "Mission Objective" nudge on youtube as well! I feel like it's one of the most effective nudges and would really make me consider twice whether to watch youtube or not.

<https://github.com/habitlab/habitlab/issues/542>

I have nudges/goals turned on, but they aren't appearing on any of the sites I enabled them on.

<https://github.com/habitlab/habitlab/issues/538>

Sidekicker, and NoComments are not working on YouTube.com. They work on "Try now" mode, but when actually expecting it to run while browsing, it does not work. I can see see Comments and Side bar.

<https://github.com/habitlab/habitlab/issues/536>

nudge "1 min assassin" that decreases to e.g. 30 sec assassin if you already spent 10 minutes on the domain

<https://github.com/habitlab/habitlab/issues/533>  
have multiple slots for work times. eg : 8h00 12h00 and 14h00 18h00

<https://github.com/habitlab/habitlab/issues/530>  
In the selector for picking work days to be active it's very ambiguous which color means on and which means off.

<https://github.com/habitlab/habitlab/issues/529>  
It seems like I can't add my own nudge idea because the "site which this nudge will be used on" dropdown is empty.

<https://github.com/habitlab/habitlab/issues/527>  
Copy-pasted the js source for a built-in nudge into the editor and ran it. Had unexpected bugs not present in the built-in version.

Copy-pasted habitlab/src/interventions/youtube/remove\_sidebar\_links/frontend.js at commit da2cf17 into the nudge editor and ran it. All the comments as well as the tile of the video get grayed out, left menu no longer shows up when clicking the button (three parallel horizontal bars on the top left corner). This doesn't happen when running the built-in nudge. (Running Chrome. Youtube in dark-theme mode.)

Shouldn't they behave the same?

<https://github.com/habitlab/habitlab/issues/522>  
I want to be able to temporarily turn off nudges for a particular website. For example, I'm watching educational youtube videos, but still want to avoid other sites.

<https://github.com/habitlab/habitlab/issues/519>

I've only been trying this out for about ~10 minutes and so far everything is working very well. I'm extremely pleased with it. However, I have different nudges enabled for different sites, but some sites are showing me nudges that should be disabled for that site. For example -- I have the more hardcore nudges like 1-min assassin and scroll blocker (cant remember specific name, sorry) turned on for both Instagram and Facebook, but only the ones like the timer/supervisor/objective nudges for Reddit. I use Reddit for work-related things often so I just need to be reminded to stay on task moreso than being blocked from the website altogether. I just opened reddit and the scroll-blocker nudge is what popped up even though it's specifically disabled for that site. Does HabitLab need to cycle through all the nudges once before it starts to show only enabled ones, or am I missing something? I can simply bypass the nudge but that defeats the purpose of enabling the other, more useful ones on this site. Thank you!

<https://github.com/habitlab/habitlab/issues/517>

When I turn off for the day/for the rest of the visit, it would be nice not to have a modal to confirm : I need one more click to close it, and it's annoying. If I wan't to turn it off, then there is no use anymore to slow down my use of thoses websites. See <https://modalzmodalzmodalz.com/> for help

<https://github.com/habitlab/habitlab/issues/516>

Even when switching the Scroll Freeze off, the feature is applied. It freezes the scroll on websites that aren't even restricted.

<https://github.com/habitlab/habitlab/issues/515>

Intervention editor does not deploy nudges on "run this nudge" button

<https://github.com/habitlab/habitlab/issues/514>

Facebook nudges are not being deployed

<https://github.com/habitlab/habitlab/issues/513>

The Bouncer nudge shows the "how much time do you want to spend on this site" message too briefly for me to make a selection. Then it moves on to the site and closes it (because the default time setting is 0 minutes).

<https://github.com/habitlab/habitlab/issues/512>  
Start making font (and other content) fade to grey with each scroll.  
Slowly the font will become tougher and tougher to read.

<https://github.com/habitlab/habitlab/issues/511>  
This is one of the most useful extensions when it comes to fighting web addiction. However, many of the nudges, such as the news eater for Facebook fail to work at some occasions. And most of them does not work when combined with other nudges. This often defeats the purpose of the extension entirely. Otherwise, I love the idea of having nudges, the pie chart for an overview and setting daily limits. In the meantime however, I will use News Feed Eradicator for Facebook, WasteNoTime and RescueTime instead.

<https://github.com/habitlab/habitlab/issues/508>  
I don't want to see the timer counting how long I've been on sites that I don't wish to spend less time on

<https://github.com/habitlab/habitlab/issues/507>  
If I can disable the activated nudges easily, then there is no need for this extension. changing habits must be forced by rules made by the user.

<https://github.com/habitlab/habitlab/issues/504>  
My settings are lost once I close Chrome.

<https://github.com/habitlab/habitlab/issues/503>  
signed in to synch, but after sign in synch not implemented

<https://github.com/habitlab/habitlab/issues/499>

Hello , I really enjoyed using HabitLab , however I noticed it is consuming 16 -> 21% CPU, even when it is toggled 'Off' or otherwise not in use while on. This is unacceptable so it is being removed from my system (MacOS 10.13.6)

<https://github.com/habitlab/habitlab/issues/498>  
XXXXXX

<https://github.com/habitlab/habitlab/issues/497>  
I no longer see the pie chart with time spent on each site. How do I get that back?

<https://github.com/habitlab/habitlab/issues/495>  
The piechart is not showing

<https://github.com/habitlab/habitlab/issues/494>  
For some reason , graph doesn't show.

<https://github.com/habitlab/habitlab/issues/492>  
there is something wrong ,the site isn't tracking my activities

<https://github.com/habitlab/habitlab/issues/491>  
i can't see my most visited sites , it's not working

<https://github.com/habitlab/habitlab/issues/490>  
I can't see the pie chart on the extension or the web page.

<https://github.com/habitlab/habitlab/issues/488>  
i need the same software for firefox , how can i do ? can you help me ?  
and send me an email ?

<https://github.com/habitlab/habitlab/issues/485>  
the extension for chrome(my verison is :Version 68.0.3440.106 (Official Build) (64-bit)) keeps hiding my you tube videos , it seems like a bug due to still keeping the names of the pages on the screen .

<https://github.com/habitlab/habitlab/issues/482>

Sometimes input messages don't work. The Mission Control nudge will display without asking for the objective (reading "Objective:" at the top), and the Bouncer will display without asking for the time, defaulting to zero minutes and immediately displaying the "Close tab / Turn off Bouncer / Cheat for 30 seconds" choice. Doesn't seem to have occurred on any of the default recommended sites, only on sites manually added by URL or the extension icon.

<https://github.com/habitlab/habitlab/issues/480>

Instead of Scroll Freezer I would just love a permanent feed remover! Let the feed be only as long as the window instead of infinite.

<https://github.com/habitlab/habitlab/issues/472>

I have an idea for a nudge. I would like to be able to remove "likes" from Instagram so that I don't see the number of "likes" I receive --only that people liked it or did not.

<https://github.com/habitlab/habitlab/issues/471>

why does it show that i am spending time on facebook when my account has been suspended.

<https://github.com/habitlab/habitlab/issues/470>

I'm really excited about using this site but I did notice that the voting at the bottom of the habitlab chrome extension seems against the purpose of becoming less addicted to sites. What's to stop people from constantly voting about the best nudges other than adding habitlab itself, if possible to do, to the list of sites to spend less time on? Just seems a bit contradictory. I like the idea of asking for peoples' opinion but I'd say cap it at maybe 2 or 3 votes per time using the extension.

<https://github.com/habitlab/habitlab/issues/469>

"bouncer" doesn't work when you open internal links from within a blocked page (such as a facebook comment thread from your notifications, a reddit crosspost, etc). WOULD be great if it worked!

<https://github.com/habitlab/habitlab/issues/468>

There should be the ability to white list specific website to encourage certain behavior.

<https://github.com/habitlab/habitlab/issues/466>

Too many notifications interrupt my thought processes. Hiding clickbait? I'm all for that.

<https://github.com/habitlab/habitlab/issues/465>

Show Stats from Yesterday (or more days from the past)

<https://github.com/habitlab/habitlab/issues/464>

not working on feedly.com -

<https://github.com/habitlab/habitlab/issues/460>

(from geza) font is off in the goal suggestion interface

<https://github.com/habitlab/habitlab/issues/452>

I don't see any option for easily changing over to purely tracking purposes - I have to manually turn off (permanently) each nudge AS IT HAPPENS in order to do this, and since this doesn't integrate across instances of chrome (all my machines), I think I'm going to have to do this for each location that I use Chrome in.

<https://github.com/habitlab/habitlab/issues/451>

(from geza) css for the suggest goals interface seems to be off here compared to other sites. shadow dom issue? also the gitter api seems to have changed breaking our report bug functionality

<https://github.com/habitlab/habitlab/issues/446>

I turned the restrict your time nudge off because it caps at 5 min. I like the idea, but it needs to be free form.

<https://github.com/habitlab/habitlab/issues/444>

When given prompts for which nudges I think would be better, I got presented with two identical options – write some code that skips the comparison when that happens? (Also, the widening of this feedback rectangle while typing is a little disorienting; fixed-width would probably look nicer.)

<https://github.com/habitlab/habitlab/issues/442>

Please add a 1 hour or 2 hour disable for the interventions, "rest of today" produces unwanted results if I click it at (for example) 1AM, I want it to come back on after a while but not 24 whole hours

<https://github.com/habitlab/habitlab/issues/438>

Is there an option to block the site completely?

<https://github.com/habitlab/habitlab/issues/437>

I wish there was an option to set a goal and get pop-ups etc just for using Chrome at all, regardless of the site.

<https://github.com/habitlab/habitlab/issues/436>

Prompts if you would like to watch again or not when using the loop feature. Very annoying when playing music in the background.

<https://github.com/habitlab/habitlab/issues/431>

how about a metric that would total the time online from various sites and compare that against the aggregate "budget of time" for all sites?

<https://github.com/habitlab/habitlab/issues/429>

It would be great to have positive nudges as well. For example, a screen that asks if you would like to spend time on a site related to something one intends to do more of (meditation, exercise, job search)

<https://github.com/habitlab/habitlab/issues/428>

This app is not tracking my activity anymore

<https://github.com/habitlab/habitlab/issues/427>

I want to spend LESS time on the nytimes

<https://github.com/habitlab/habitlab/issues/416>

There was something to remind me to visit important sites , , where is  
that feature ?

<https://github.com/habitlab/habitlab/issues/415>

(from geza) regression in youtube pause videos intervention

<https://github.com/habitlab/habitlab/issues/411>

I'm trying to set the hours nudges are active via "only work hours." It  
won't let me set them also for my sleep hours! I would like the  
nudges active from 11pm to 5pm

<https://github.com/habitlab/habitlab/issues/406>

YOU GUYS ARE AWESOME! I love nudges. I love Stanford. This helps me so  
much. I wish everybody knew about it.

<https://github.com/habitlab/habitlab/issues/404>

Guys, I fucking love what you made! I send you all my gratitude for  
your wonderful product!

<https://github.com/habitlab/habitlab/issues/399>

Settings aren't synced between browsers

I installed the extension on Chrome and customized some settings. When I  
opened Canary Chrome, the extension was installed , but the settings  
weren't synced. It's possible that this is a problem with Canary ,  
but my guess is that settings also aren't synced between Google  
Chrome on different desktop machines .

The ideal solution would be to have user accounts , so that once the extension is available on other browsers , your settings could be synced between them. But it would probably be much easier to rely on Chrome's syncing for the time being , which would provide most of the benefit. (People probably tend to use the same browser on different devices.)

Keywords: sync , synch , synchronize , synchronization

<https://github.com/habitlab/habitlab/issues/395>

No increase in time though I often visit the page I want to reduce time on

<https://github.com/habitlab/habitlab/issues/389>

You're telling me that I spent no time on FB or Youtube , when I definitely did. I've not turned off HabitLab , so how does your measurement system work? Are there any other Chrome extensions it doesn't work alongside with?

<https://github.com/habitlab/habitlab/issues/386>

I use lots of PCs and MACs and would like to see my HabitLab a rather cross-device , Gmail-based account. Thanks a lot for the great work.

<https://github.com/habitlab/habitlab/issues/385>

turned on for twitter - minute watch distracting / annoying , so turned on (?) "supervisor" , turned off minute watch. It doesn't take. It still annoys. I'm about to simply turn off habit lab as a result .

<https://github.com/habitlab/habitlab/issues/384>

tiny bug, every midnight it shows a popup about having accomplished my goals of the starting day which obviously of course not the case

<https://github.com/habitlab/habitlab/issues/381>

I've had the extension installed for a few days now, but so far it hasn't tracked ANY minutes on ANY website. Happy to provide a list of current/active extensions if it would help. My first instinct is perhaps "HTTPS Everywhere"?

<https://github.com/habitlab/habitlab/issues/380>  
Timer stops sometimes

<https://github.com/habitlab/habitlab/issues/379>  
It does not accurately track the amount of time spent on Netflix, unless you are moving your mouse around. It starts counting, but then pauses counting after your mouse hasn't moved for a few minutes - i.e. passively watching a video. It was unable to accurately track the minutes spent even though the timer function was on, and I was interacting with it by pressing "snooze" a few times.

<https://github.com/habitlab/habitlab/issues/377>  
As an alternate to a addictive site, you should let people completely step away and engage elsewhere. The alternative to a website should not only be another website.

<https://github.com/habitlab/habitlab/issues/376>  
I hope this add-on works with Chrome on incognito windows as well!

<https://github.com/habitlab/habitlab/issues/369>  
At times I am using youtube for educational content and would be nice to submit pages as educational so they dont trigger any HabitLabbehavior.

<https://github.com/habitlab/habitlab/issues/363>  
It appears that the clock does not accurately count time spent on Netflix when in full-screen mode. It begins counting, but then pauses.

<https://github.com/habitlab/habitlab/issues/362>

I'm loving HabitLab so far. An additional suggestion: I want HabitLab to play a distinctive sound when a nudge is activated. Noises can be annoying, but they can also be a powerful part of retraining our habits. The user should obviously have the option to silence HabitLab, but the sound option would be a big upgrade. I would be glad to help develop a sound (or array of sounds) that could be used for this.

<https://github.com/habitlab/habitlab/issues/361>

Please allow option to not turn off buttons, e.g. "turn off feed eater button". Ideally this is a global option across all extensions.  
Thanks! :)

<https://github.com/habitlab/habitlab/issues/348>

There should be a way to combine sites into the same grouping (i.e. smile.amazon.com and www.amazon.com, and to account for different parts of the site, such as music vs. shopping)

## Appendix F

# Responses to Uninstall Survey

The following are the free-form responses to our uninstallation survey. Some earlier feedback items may be missing, as we accidentally lost them during a server migration. We also removed or modified items that revealed the submitter's identity or personal information.

- 1: It did not work (nothing was registered)
- 2: another testing
- 3: Times logged don't accurately represent engagement with the site , e.g . playing a video in the background while viewing a different tab didn't count toward time but leaving the youtube home page open and idle did .
- 4: Neighbor installed this to my computer without my consent!
- 5: because i want to
- 6: muy molesto
- 7: Looked like some nice options for streamlining sites , but it 's actually a nanny. I don't need a nanny whining at me.
- 8: I reached my goal to reduce time spent on certain pages. Thanks folks !

- 9: Didn't seem what I was expected. Installed two minutes ago and removed it.
- 10: I wanted something to track all websites, rather than having to opt-in to each.
- 11: I was just worried, I mean it's standford and all, but a bunch of students tracking everything I do and all my browser history, that just felt too much of a price to pay
- 12: ';
- 13: I forgot that I already had another program that did essentially the same thing for the computer in general instead of just for this particular browser
- 14: ...
- 15: changed my mind
- 16: Translate in french please
- 17: surveys done
- 18: You covered up useful buttons. Don't do that.
- 19: Installed for a study, now the study is over.
- 20: just wanted to track time spend on youtube
- 21: it wasn't working on tumblr.com
- 22: I noticed that I was going on imgur, youtube, facebook (my choice of addictive sites) more, after I had installed the extension. So, I'm uninstalling. I think the extension made me more conscious of the fact that I was visiting the sites, but maybe the rewards were making me go back to the site? I'm not sure.

23: This is Matt, just deleting and reinstalling for something

24: Look and feel weren't great, I feel like it annoys more than it helps.

25: I don't need it :P

26: Wasn't sure if it is effecting battery life

27: Setting page was continuously popping up despite filling the setting

Settings kept on reseting to default, despite setting the facebook specific timers they kept on getting disabled.

This was causing way more distractions than facebook which is sad.

28: it was just annoying you out of not using sites, not convincing you to. It became like ads, they are always there. But you don't like them and turn them off with ad-block.

29: I should be 0 this time

30: Hello ,

This is the web browser for my raspberry pi 3 b. I use website such as youtube and reddit for information and programming advice etc e.g terminal and python info. I do not wish to use less of these websites while doing this. However on my phone and laptop I do want to use less of certain site , as I use those devices for socialising and fun but it has become rather consuming of my time. BTW Habitlab is great application I hope the team achieves what it set out to do.

Thank you kindly

Regards

Chris

31: I don't spend a lot of time on facebook or other social media.

32: Was awesome, but was making chrome really slow, i mean really slow!  
Seems like you need to fix some memory issues.

33: don't need, just wanted to try it  
thank you

34: I liked the interventions but not on every page change or load , that was just a bit too much. Also I couldn 't work out how to disable the thumbs up gif that appeared when I closed a site I was trying to reduce use of. Now that was annoying. Otherwise though it 's a cracking extension! Good stuff.

35: Just testing

36: does not fucking work, hiding the newsfeed in particular

37: I was bored with it.

38: Did not turn off when requested. Kept on nagging even when disabled.

39: Found out I have good habits already.

40: I often browse incognito and it does not pick that up. Not that it should.



"Click here to finish setting up HabitLab"  
"Click here to finish setting up HabitLab"

42: i don't like this

43: i love this

44: Debugging (am at Stanford)

45: They are preventing me from adding youtube songs to my youtube playlists.

46: testing to see if server is up

47: Only need on my main computer, and way too many little notifications in the bottom right corner with no clear easy way to turn them off

48: Extensions worked for only 5 minutes (until I started to modify the settings).

It's now completer borked. Clicking on "Options" sends the browser in an infinite loop.

Too bad, this looked like a useful project.

49: no like interference in all

50: There was no way to turn off notifications in chrome

51: It does not have any auto sync features. because I use Multiple OSes  
that's why I'm having so much problem.

Please add auto sync features of Habits.

52: Just was testing the app

53: I wanted a timer for every "domain", it can be good for statistics  
of time..

54: gyguv

55: may cause lag

56: Catastrophic stability problems after installation; may be due to a  
different extension

57: Couldn't figure out how to turn off the "clapping" gifs that appear  
when leaving a website.

58: Dont need it on work computer, just at home

59: downloaded twice accidentally

60: this is my wasting time computerand i dont need it

61: would not turn off even after I "turned this off for today" when I  
tried and corrected it it told me it was already off. Also was set  
to only run during work hours, and never stopped on the evenings or  
weekends. Basically it did nothing but bother me when i wanted to be  
online, and never helped me curb the times i didn't. So utterly  
useless!

62: d

63: A cool concept! Unfortunately, I don't use Chrome that much so it is not very useful for me. A mobile application would work much better for me since I tend to procrastinate on my phone.

64: There was a very irritating bug with Klout.

65: I love the idea but I'd like stronger nudges and better tracking of the time spent on each site.

66: I was interested in tracking my usage to start, instead of setting interventions that I may not actually be concerned about. Additionally, much of my "wasted time" usage is on mobile which this wouldn't help me track.

67: Made Facebook unusable. Which might be the point?

68: Someone else installed. Did not want it

69: Gave up

70: LAG.....

71: i'm chopping and changing computers too much...

72: The "goals achieved" notification was annoying, I disabled it and it still intervened

73: n

74: hjhjk

75: .

76: only test install, good apps guys, regards from spain wihout user anonymous from castila la mancha :)

77: Could not edit settings properly. Even though I removed Gmail from sites to avoid, I still got notifications.

78: Should be able to monitor all websites first automatically, and not only user defined sites.

79: Testing

80: i don't need it

81: I think it messes with the Google Play Music

82:aaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaaa

83: Made me realize I don't have Facebook addiction, spending less than 30 minutes of my desktop time on it per day.

84: ./'1

85: Would have preferred a gentle log - perhaps emailed - giving usage statistics. In the present form this operates like pop-up ads. Still, it was reasonably insightful into understanding my usage for the time I used it. Thanks.

86: So

87: Seeing if this extension is causing gmail compatibility issues

88: Creating an intervention (such as temp\_intervention created in the tutorial) does not seem to work at all. I'm reinstalling immediately to see if that fixes it.

89: I prefer the "Forest" application.

90: kept telling me 8/8 achievements complete when in reality I had not.

91: For some reasons Facebook blocked me and I am trying to figure out the reasons. What I know that the Hitlab extension was deactivated (not by me) and then, boom, I was blocked.

92: Software was buggy

93: all good

94: did not appreciate the clapping people. had issues with google music and now checking if an extension is the cause

95: not sure I really needed it – mostly use facebook and youtube for technical research and discussion. Good idea, though

96: disturbing gifs

97: I just don't use my laptop as much as I thought would be necessary for an intervention.

98: Not interested

99: I rarely waste time on my desktop. This would be much more useful on my mobile phone.

100: not what i was looking for in software.

101: Interventions are not forceful enough. They are too easy to click around or disable.

102: realized not spending as much time on websites as i thought...

103:

104: Uses a lot of CPU and memory without doing anything

105: Mostly it was the bar covering up facebook message indicators.

106: Unable to change settings that appear in startup. I couldn't find a way to set one type of nudge for all websites. It is really annoying to change settings for every single website. Hence reinstalling it.

One good feature will be to create a group of websites with similar nudge settings!

107: settings changed on me all of a sudden with caue and i got sick of going back over settings over and over to turn things off.

108: I like all or nothing.

109: It's quite possible something else was causing lag -- but lag there was. I also was just checking it out. I don't really use facebook or youtube.

110: I figured out that I actually don't spend as much time as I thought I was. I thought I was watching videos for 4 to 5 hours a day, but the timer stated only watch them for 30 minutes.

111: Not only annoying but they cover up important part of the screen

112: sucks

113: Cause Facebook news feed to constantly refresh itself so I was unable to read or click any links there.

114: I used the extension to curb my Facebook habit and eventually gave up Facebook altogether - something I've been wanting to do for a long time. Thank you.

115: no need

116: Don't need it

117: I do not use it anymore

118: non lo voglio

119: My youtube doesn't work becauseof it

120: battery use

121: I love the app - I'm just removing temporarily to see if it's affecting another app (Freedom.to)

Thanks

122: I 'm weak....

123: tuke

124: I 'll be back soon

125: I just didn't understand the concept before downloading and it 's intentions aren't my demons as it happens.

126: became unusefull for me;  
good initiative any how

127: NA

128: Extension did not work as expected and I was experiencing bugs

129: I want to try other types of chrome extensions to block time-consuming websites and don't want to mess with your data .

130: performance... seems to use a lot of CPU all the time...

131: THis extension was causing system wide lag

132: I use social media for work - but your app is great and I will keep it running on my personal accounts

133: Por enquanto n o preciso desse tipo de ferramenta no meu navegador .

134: Daily results show 0 time spent for almost all sites , despite that definitely not being the case .

135: excessive battery consumption

136: All it took was the fact that settings don't sync , so I would have to reconfigure for each instance of Chrome .

137: Very Good , but want to reduce installed extensions .  
Might install again later .

138: Spend too much time on the relevant platforms for work purposes .

139: dcsdf

140: Blocked usage of 1Password extension , which is critical .

141: 1

142: lags and bugs

143: Time spent differs drastically

144: wanted to install it on my private profile , not my work one .

145: just checking it out

146: dssd

147: I think I misunderstood the point of the program .

148: The Youtube sidebar still loaded .

149: didn't really need it. Turns out... I don't really binge watch after all.

150: No use

151: I need something mor physical, i can be distracting out of my computer

152: ijgougooughu

153: Consumes too much CPU

154: On Facebook, I couldn't click on regular posts (not just clickbait posts) that I truly wanted to see. So I disabled the clickbait feature and it was still happening. Then I disabled the whole app, but it's STILL happening. So now I have to delete the whole app to see if this fixes the problem :(

155: cxsd

156: nnfgnfgnngfnfndrsxnrtsynr.shrehmsortihmnsrth,rtesmhposritmh[oimaerg][ognm[09trignjm3io[54ny9ui35y-

157: Not effective for me – the interventions just slowed down my procrastination and made it take longer unfortunately! :)

158: its stupid asl

159: Basically just irritating.

160: I like what you are doing and see a great purpose. but when I was on certain pages the pop up from habit lab was in the way of seeing what I wanted to look at and then I felt frustrated and somewhat invaded..... due to the fact that I thought I had shut it down and then somehow it was on again.

perhaps there is a quick habit blurb.... about how to shift when you get stuck on the computer etc, or using triggers to support positive habits or kaizen ,

If it is purely research base maybe there needs to be finessing around how the habitlab pops up or the look of it, it's placement on the site..... right now it feels slightly naggy, perhaps it could feel supportive in the look and placement of it. Like I said .... I like what you are doing.... It's super smart and hopefully will help gather the data you are looking for.

161: Didn't seem to work

162: Kept me off Facebook , but was not smart enough of a tool to know when I needed to check a message versus just idling

163: Used it over my study period , don't need it anymore. Thank you for it !

164: gmail and youtube stopped loading

165: seems cool but i feel its too intrusive , probably helps others who sit on facebook all day

166: Interfers with printing emails -- pops up as if one wants to print it

167: reinstalling so i can see the habitlab panel , not finding how to access this normally?

168: I was hoping for an analysis of my internet activity. This seems to be designed solely for the purpose of reminding users of youtube and facebook time .

169: Does not remove youtube sidebar and comments section

170: Caused interruptions when listening to youtube in the background.  
Many videos would randomly give an error and refused to be played about 30 or so seconds in.

171: I achieved my goals

172: It gave me the info I needed

173: it reached its goal as I am more cautious about my net usage

174: Didn't catch any website anymore

175: need to reinstall

176: Needs to synchronize stats between machines , ineffective otherwise .

177: I deleted the wrong one

178: Installed , because I was curious . I don't have issues with wasting time in the Internet .

179: i dont actually spend too much time anywhere so after trying it for a few months (i think?) its rly just annoying . but if you do have an issue with going on facebook or reddit (or whatever) too much , i'm sure its very helpful . if that ever happens i'll reinstall it :)

180: On one of my other synced devices (RCA Tablet 10.1 w/windows 10 new April 2018 ver.) it keep crashing and literally giving me a Black Screen of Death . It would just go blank and one time it did ask if I wanted to submit a crash report all other times the error was not scene just would not work .

181: Wasn't using it

182: fuck useless , useless team

do better things with your life

183: xzy

184: I clean toolbar and delete all extension

185: I feel it does not effective.  
Rescuetime works better for me.

186: only need on work computer

187: a

188: No instructions on how to use it.

189: Was not effective at all. Provides a false sense that something is being done, making the problem worse. StayFocusd will do well to block these sites in their entirety rather than try to alter the design that tech giants spend billions perfecting.

190: ;not finals anymore!

191: wr

192: I wanted to crush my laptop's screen every time a stupid high five you saved a minute fucking awful meme was making an appearance cool marketing of extension but treating me like a stupid mindless kid is beyond annoying .

193: I needed to get rid of the Ram Chrome was using.

194: interventions overlay makes webpages unusable

195: I couldn't find how to de-select some (already selected) web pages.

196: Ticker blocked screen content

197: i will be back

198: hgh

199: ram usage

200: Using HabitLab in Chromium on a linux operating system , the browser became slow and unresponsive .

201: Didn't really need it

202: Those "nudge" ads were too annoying

203: It was marking work sites as time wasters . It made the numbers very misleading

204: was using multiple extensions fuck extensions they open up like 9 processes on my compute that pull 24k memory each

205: I wish you could be more interactive with the nudges/interventions , they felt preset without much control over them

206: After I disable it , it continue to be active while I was working on my research paper. I thought it will help me when I spend to much time on social media sites , but not when I was on google researching or writing .

It is good , but not for me.

Thanks

207: Sometimes I actually spent more time on the site because of the content obscured by the intervention , otherwise did not see habits change .

208: Added the extension at work . Need it at home .

209: Love the idea , but genuinely don't go on Facebook or Youtube or Amazon

210: It 's Drew :-)

211: Hootsuite won't connect with FB but it 'll connect with Twitter and LinkedIn . I think HabitLab might be the reasonl.

212: I used to enjoy the app but I'm not a huge fan of the new update in which I receive constant prompts on tabs that I'm using for work ( Dropbox etc). Even though I clicked 'no reminders for this site ' it continued to pop up.

213: it stopped working time ago

214: Wasn't working when I said to disable .

215: erfdcvgre

216: ffffffffffffffrr

217:

218: doesnt work with incognito mode

219: I need to free up ram.... otherwise the extension is great!

220: In the beginning seems to work with Firefox but crash then :/

221: I set it to 15 mins and then opened three links from the same site ... each tab gave me a new annoying warning and then the third one told me "time 's up" and tried to turn off the site . Also it 's super confusing which button to click to make the warnings go away for my damn 15 mins . Uninstalling because this is just too annoying when I just wanted a simple 15 min timer .

222: I do not need that .

223: fdih

224: I have my habits under control for now! Thanks!

225: Wanted to turn on all nudges at the same time

226: can't tell if it's available in incognitio mode

227: If I want a nudge , I WANT A NUDGE. PLEASE STOP DISABLING THE ONE I WANT TO SHOW ME SOME OTHERS, I WANTED THAT ONE

228: The nudges change all the time , and I would like to choose definitely which to apply .

229: Would not stop blocking sites when app was off on "turn off for the day"

230: Didn't need it any more thanks!

231: Don't need it anymore

232: Android initial automated setup buggy. couldn't proceed out of ' display over other apps' step . maybe will try again with later version .

233: Said no nudges were active even though they were

234: The timer was incorrect (way below actual value) , making most of the interventions useless

235: Disabling interventions didn't stick , specifically "Bouncer".

236:

237: going to purchase new laptop

238: Brakes youtube

239:

240: I enjoyed this -- it was very helpful for a month or two, but after that I felt like I'd gotten most of what I would get out of it. Now I don't have as much trouble managing my time and I can just use a basic site blocker if I'm going to some site too much. Thank you for improving my life!

241: Prompts to try new interventions were annoying

242: lost youtube thumbnails

243: x

244: xxxxx

245: Had it installed on work computer, and needed to remove it to comply with extension installation policy.

246: Broke YouTube, and I don't really need it on desktop.

247: Slowed internet down

248: Causing so many problems than advantages

249: Great product. Just updating some settings.

250: didn't use much.

251: Glitchy. The turn off button did not work half the time and other buttons did not work.

252: n/a

253: the extension is still primitive and buggy... it shows at midnight a notification that I've accomplished my goal while it's not exact it's just notifying from wrong data about the upcoming day not the previous one...

254: way too much lag

255: Interventions were annoying. I will still try to control my browsing habits. I wanted it to only work from 8:00 A.M. -- 5:00 P.M., but when I try that, the interventions still go all of the time  
...

256: The extension was preventing youtube thumbnails from loading correctly

257: Some assets aren't loading on youtube and other sites.

258: I really like everything about this app, but I despise how often it asks me how it's doing. I set the settings as I wanted them and don't want constant popups asking me additional things.

259: I don't find myself needing it in my new job.

260: Starting to break in numerous areas on Chrome, such interventions showing up that aren't selected, broken websites (e.g. Youtube thumbnails).

261: Need youtube for work

262: Was not working correctly

263: asdfasdf

264: high CPU usage

265: Because you redirect to this fucking page when I say I'm done with you. Stop being a dick.

266: Turns out I don't actually waste that much time on specific sites!

267: I did kind of like seeing how long I had been on gmail , but in general it wasn't very long , so I wasn't getting feedback that really gave me insight into where my time is poorly spent. I also don't use FB and twitter much. I could see in the activity monitor that a chrome helper was taking up way too much bandwidth and killed it , and it became clear it was related to the habitlab .

268: Doesn't works correctly

269: bl

270: Everything was fine with the app , I liked it. I'm just not interested in using it anymore because I don't have work to do on my computer.

271: no more need

272: Constant CPU consumption ...

273: If you could link all browsers and devices used throughout the day , then it would be better .

274: rther h rthe gh rtjhertjer ej

275: rther h rthe gh rtjhertjer ejrther h rthe gh rtjhertjer ejrther h  
rthe gh rtjhertjer ejrther h rthe gh rtjhertjer ejrther h rthe gh  
rtjhertjer ejrther h rthe gh rtjhertjer ejrther h rthe gh rtjhertjer  
ejrther h rthe gh rtjhertjer ejrther h rthe gh rtjhertjer ejrther h  
rthe gh rtjhertjer ej

276: Was inconsistent as well

277: I felt that I would need to put in some time to learn how to use it , and not knowing how to use it the interventions just seemed annoying. Maybe if there were a two minute video I could watch that said "to set it up do these three things" or something like that. I guess that is a round about way of saying that it feels too complicated.

What I really want is to have a row of timers on the top of the screen that show me how much time I have spent on this website today , this week , this month , and the same for all websites that I consider time wasters (facebook , news sites , twitter) . It would be really nice to be able to go to a page that shows the times for all websites and I can group them together e.g. bank and credit cards (to answer the question "how much time do I spend managing my finances ") etc .

278: I downloaded this app as an experiment. Unfortunately , I work in an advertising agency , and being rather social inclined is a solid part of my job . Hence , the need for social media .

279: I tested out of curiosity . No real need .

280: Sites I told Habitlab to ignore would continually show pop-ups . Not worth the hassle .

281: Thanks. i instaled it because i read a review and [raises of it . Wanted to expeience and hence insalled . Now i t

282: Thanks. i installed it because i read a review and [raises of it . Wanted to experience and hence installed . Now it is enough .

283: Checked to see what it is and found it is not useful for me

284: Malfunctioning intervention updates

285: . Feedback

286: Constant 16–21% CPU usage while the browser is open, even while HabitLab is toggled 'Off'

287: It wasn't working properly, I had set many sites to 1 minute assassin, and it did not work at all

288: got tired of it. It's an awesome app but not for this moment

289: Just wasn't using it anymore

290: It really worked for me during the time I needed to sit down and do very boring chores on my laptop. Now those boring chores (aka writing my dissertation) is done, I am ready to remove this app and indulge myself in a little bit of web-surfing. I will reinstall it if my self-control is out of wack.

291: Decided to simply block the site using Block Sites

292: Did not store settings between launcher restarts

293: Don't use it nowadays

294: was not showing pie chart and other data .

295: I just wanted to start with a simple background timer for the pages i visit and go from there, but its so hard to disable everything. Very annoying.

296: bla bla bla

297: I haven't taken the time to set it up to not be annoying. I may reinstall at some point! I was really just trying it out to see how it might help some of my students..

298: not intuitive to have the banner appearing the minute i start using the internet – this should focus on as of a specific time or when more than x time is used on one site and as of then present the user of most frequently accessed sites , length of time on them and suggestions on when to set alerts

299: Would be nice to have the option to choose which kind of notification you are given vs giving feedback on what was just done. As well as move the notification to a side of the screen that's not blocking what you're currently doing (timers that stay up longer ). Otherwise it helped me a lot to curb my time wasting online . I feel I can manage on my own for now but will reinstall if it starts to get out of hand again . Thanks !

300: On few computers I had to decided EACH AND EVERY TIME WHAT I WANT ... bit I decided already what I want... it was extremely DISTRACTING :(

301: I decided to stop using it .

302: not interested

303: h

304: Loved it , but was causing crippling lag whenever I had a few tracked tabs open. Maybe I 'm imagining it , but when I split my browser into multiple windows , it became unusable .

305: Adding an option to completely block websites without a way to get around it for a set period of time would be great

306: Unfortunately I felt likely the nudges were creating more dependence on FB and other social media

307: Was sometimes forced to answer what I thought about the nudges .

308: Keeps crashing my browser .

309: An opt in for sending anonymous feedback would have been appreciated

310: put in a switch off button

311: confuse !!!!!

312: Worked well, don't need it anymore.

313: The nudges were completely ineffective and ended up being more intrusive than anything.

314: Sorry, but it's a shit. It constantly overwrites the settings, switches on modules that have never been plugged in. It would be good. Bye bye.

315: I was just looking into whether it was blocked behind my employer's firewall. It's not. Thanks!

316: interventions felt a little too lax. also conflicted with another extension I was trying out, Prod, that I decided to keep using.

317: Timing seemed to reset when I went to another page, so the clock/measurement did not seem effective

318: Got over the digital habits I wanted to get over. Thank you!

319: ,

320: Will reinstall soon

321: aaa

322: half the interventions were really annoying and made me think the internet was down

323: It wasn't helping me to curb my web surfing. I simply became better at tolerating the irritation of the interventions.

324: Less passive than expected

325: When I've opened youtube, it hasn't shown the player.

326: Absolute annoying piece of shit

327: It totally worked. It worked so much that I don't even need this app anymore ;) I no longer spend more than 5 minutes a day on Facebook and other distractions. Thank youuu :)

328: It was just driving me crazy. And I couldn't find out how to get rid of it. I thought it would measure instead it bothers.

I didn't want to get rid of it, I wanted to turn it off, maybe a gentle email in a month or a year when it would gently email me or something saying it has been improved. You know?

329: Use a lot of processing power

330: I didn't notice the sniper one often and I would be in the middle of a message to my wife and it would close it so I would have to start over again.

331: I really liked this tool. I think it actually worked to create some good habits. But it's slowing my system down.

332: i dont like it

333: fu

334: no thing

335: was great and helped me out of facebook.  
As i am out now - no need for Habitlab anymore, thanks !!!

336: z

337: Did not see anything changed with only 1 nudge per page active. And yes, i had goals set!

338: I installed it on the wrong browser, going to reinstall in the right one

339: i didn't use that so much

340: 44

341: the banner would be on top of youtube search bar, thats it

342: I don't need to know I've been on Youtube every 2 minutes while watching videos. That's obnoxious

343: soy un aidcto

344: ,khjb

345: I don't use for a long time. I can control myself.

346: bvcbc

347: displaying the habitlab goals was distracting

348: Going to give away this machine

349: I was having a problem and wanted to see if HabitLab was contributing to it

350: I need to reduce my online reading in general, without naming specific sites (there are too many).

- 351: I'm reinstalling it to see if it fixes the bug : nudges would not activate ...
- 352: Being unable to see a history of my behavior was a big factor in my removing this feature.
- 353: No longer interested in the help.
- 354: Prevented work as a web developer. Stopped cards opening - "Detected Link"
- 355: I think I'm now in control of my habits , thanks!
- 356: caused multiple websites to malfunction. i'll most likely try re-installing it.
- 357: interventions didn't make me use the sites more. It just irritated me.
- 358: I'm not sure if this extention collecting my information or not so I think the best way is uninstall
- 359: Breaks everything without any reason
- 360: seems buggy after updates. uninstalling and reinstalling usually helps .
- 361: It was buggy
- 362: I was testing it to recommend to students. Interventions were not necessary for me, personally .
- 363: No longer needed
- 364: bye

365: Your messages are intrusive and hard to remove/clear from the screen. The resulting annoyance defeats the purpose of having this extension .... Please work on this.

366: '  
/

367: w

368: add a feature to turn it off for more than a day

369: the placement of the timers and text box is annoy, sometimes it blocks the search box etc

370: I can't turn off HabitLab with Gmail while crashes continuously if the addon is running

371: I liked the interventions during periods where I was very busy, but they didn't feel flexible enough for every day use. The interventions were a bit too intrusive, and the option to "turn off" for a day often didn't work or froze the system. Still, I would consider re-installing in the future.

372: Many times i'm not able to turn off the intervention temporarilry and they keep coming back when i refresh the page.

373: fans are running at 6500rom on macbook air

374: slowing chrome

375: Just cleaning up everything after getting hot by WeKnow virus

376: I felt like all of the interventions just ended up getting in my way and didn't seem to be helping either. I think for whatever reason the interventions Habit Lab supports were not effective in changing my online browsing behaviors.

377: Bad

378: i thought the app would be more effective for my addiction

379: na

380: didn't really do anything

381: it works

382: HitLab taught me that I value almost all of the time I spend on social media and don't consider it wasteful.

383: very high cpu usage

384: Simply wasn't using it. It's great, it just wasn't for me.

385: no longer required

386: They popped up in places I needed to get to and I couldn't move them away.

387: In this period I'm just full of things to do that I barely have time to open Facebook, but also in my free time I prefer doing other things so I can say "mission accomplished"! BTW great extension, if I'll need it again I'll reinstall it for sure.

388: Bug. Would cycle through the same "medium, heavy, light" screen even when I made a selection

389: Chewing up too much of my CPU

390: you kept popping up, once I tell you I need you to count minutes, you kept asking me the same question always! You should have where one marks what he/she wants and when he/she needs the change, he can get to settingd

391: dsadas

392: The objective nudge was not working at all and the hide comments nudge was active even though i set it myself off. I'm trying to reinstall to see if it fixes itself

393: I changed my work and almost not using bad habits sites :)

394: kept asking me what level of interventions i wanted when i opened a page. when opening multiple tabs on FB i had to click "light touch" for every single tab. maybe it didnt save my setting since i have low disk space but ende dup being too annoying

395: Sorry , just couldn't figure out how to get rid of the timer at the top of Gmail. And it is particularly annoying that I was typing up homework. Couldn't bother going thru to fix it , instead I just removed this addon

396: It was preventing from doing some of my work and I do not use facebook and similar sites as such , any more.

397: Didn't need it

398: need to work with facebook and were annoyed to have it on my work computer

399: I was just testing it out because I'm interested in your research!  
- Cori Faklaris , CMU

400: I wasn't ready to fully use it. Great for a work machine , my home machine my habits are way too unproductive , yet satisfying . Didn't want to log it as annoying , as the concept is great and ill prob reinstall soon without interventions . Thank You!

401: It was really buggy . It would not remember settings .

402: Gmail could not be loaded. Got the message Error 404. The requested URL was not found on this server. Thats all we know.

403: Settings should be set only once and not asking me every time I visit a site

Too much focus on Facebook, almost useless for persons who don't use it

404: I never got the time to check it out

405: I wanted to study my habits with some figures and charts, not a nagging system to get rid of some addiction. The software is clearly feature-rich, but it is too much for my interest, and by default it is VERY intrusive and pushy. Instead of showing giant buttons blocking the content, why not just a little pop-up, that you can deal with later? cheers

406: Please, I'll be back soon

407: available websites are not an issue, need option of adding more sites

408: I don't think I need it.

409: Interventions were inconsistent. Sometimes showed, sometimes not.

410: I use a whitemode setting with Blocksite, which blocks all websites until I allow it. Also, scrolling on the setup page was a buggy mess with a lot of stuff too large to interact with.

411: Can't use in incognito tabs.

412: There currently seem to be a lot bugs that making it behave confusingly

413: Dont really need it. Just wanted to try it.

414: I've worked quite a bit to optimize my devices for minimal distraction , so although I found the tool useful for many without that discipline , this was just not for me.

415: Too much unnecessary prompts !

416: It didn't seem to register my selections so I had to make them anew each time I opened gmail.

417: my problematic use is on my iphone , not this computer

418: The extension didn't work on YouTube. None of the selected interventions seemed to occur when browsing the site . I did have the site activated in the settings .

419: 'Nudges' would not disable . If I could only use the notification / banner nudge , I would be happy , but no matter how many times /where I click 'turn off ' for Scroll Freezer , it will not disable .

420: Just wanted to purge some bad data . Left a tab on overnight , and it thought I spent all night on Youtube .

421: Heavy handed mode doesn't seem to work

422: Interventions were disruptive . I had to spend additional time to get around the interventions which meant more time was spent on the computer . All I wanted was a simple timer to track my usage .

423: I wasn't spending as much time on reddit and twitter as I thought I was . HabitLab helped me be more conscious of the time am spending , it just wasn't as much as I thought it would be .

424: this is a bug when used didnt initialize that stuff for the end

425: I think it's part of why Chrome is suddenly chewing up SO MUCH MEMORY and I can't have my computer running this slow. Interferes with work .

426: I don't seem to have a problem. My facebook use was about 11 minutes. It is a good idea though.

427: Very buggy. It decided I had reached my max for Twitter as soon as I logged on

428: New Macbook

429: Makes hangouts not load chat history

430: Makes hangouts not load chat history. Now this form doesn't close on submit.

431: I mostly just wanted to monitor my time spent on various websites , this doesn't provide as detailed of a breakdown as I'd like (think something more like screen time from iOS)

432: Would love it to be less intrusive – was expecting interventions to be more streamlined and UI-friendly – too clunky and while I chose 'light weight' you took over my whole page. It also didn't recognize cross-tab , when the domain was the same, I had to re-submit interventino notes .

433: Didn't really provide the data I was looking for

434: It was showing up on other sites I hadn't asked for , and was buggy on twitter .

435: Youtube would glitch , and HabitLab would keep popping up with Annoying Interventions , even when I disabled it. Glitchy .

436: Falta idioma espa ol , saludos desde M xico

437: I need more than top 5 most visited websites

438: I decided to use a program which can also track time spent outside of Chrome.

439: It wasn't working! I wasn't receiving any nudges on my target sites even though it was on and I had every nudge enabled.

440: While having a job now it began to be annoying having these interventions on my personal laptop.

441: m

442: i selling my laptop

443: Can't see historic trends

444: Does not get along with my work network

445: It's good, but i use youtube to listen to music while working so it's not a time waster for me.

446: /

447: re

448: all good, just realized sites I was tracking are needed for work too often

449: i need a instant cut off like a website blocker.

Ideally your tool would then show me how often i had tried to access the blocked sites

450: 68

451: 68

452: A very good idea. Nicely done. What gets measured gets done. But I'm not so much interested in creative ways to limit web-surfacing of particular sites as much as getting a handle on how much time I'm spending everywhere. There appear to be other other Chrome extensions that are better suited to my needs.

453: Not for this moment. But its very useful.

454: Full of bugs

455: Wasn't valuable for me.

456: I want to use in another time, when I get a job for instance

457: Wouldn't allow various mailboxes in Gmail to load

458:

459: Conflict with other extensions...

460: no reason. just cleaning up

461: I already clicked a box that is self explanatory. Why do I need to write stuff here?

462: didn't work

463: blah

464: 11

465: On Android, the overlay messages, once closed, would still disable the ability to interact (tap) on apps. I'd have to close them and reopen.

466: Did not feel effective

467: the time-on-gmail pop-up kept breaking my concentration.

468: asd

469: Not clear what it was doing for me, did not see a clear summary of data (maybe I didn't look?), interventions annoying.

470: malo

471: Stop eating CPU

472: Would be great if I could customize periods of usage as different segments. It's annoying to turn off nudge for every youtube tab for instance when watching youtube IS the objective.

Maybe a simple on-off button, and twice a day or something analysis of the usage, and customizing nudges based on the usage.

For example, I may be working and playing music at the back on YouTube. The nudges would override the autoplay and I'd be switching to YouTube to more often.

473: Constant high CPU usage

474: I did not need it

475: It just didn't work as well as I needed :(

476: I could not choose which nudges to activate

477: It keep asking me to choose?

478: After I started with HabitLab, I was no longer able to view lowercase X's and Y's. i.e. The message above states: "Sorr to see ou go!" Not sure if it was HabitLab or another issue, but that was the big change to my computer so I'm removing it.

479: cleaning up chrome - will reinstall later

480: I decided it was more effective to block the url from my system entirely .

481: Did not feel the need to use this

482: honestly i didnt know habitlab was to keep track of online usage only , I thought it was going to help with all Goals Settings . Such as smoking less ; Drinking more water ; keeping up with a good diet ; Etc . Good work but this app is not for me . Thanks You . I will keep you guys in mind if i ever need help with my online usage .

483: Not forceful enough

484: Buggy , especially for non-preset sites .

485: For me , a program like this would be far more beneficial on my phone than on my computer . I'm really pretty productive on my computer . I spend too much time on email , but at least that's productive work email . It's on my phone that I tend to get lost in scrolling .

486: P.S. Playing the intervention at the beginning and at the end of YouTube videos

487: please support chinese    please    support china web

488: b

489: Some of the interventions were effective , but I'm going to try a more aggressive blocking extension . Thank you for doing this work , and I wish you the best in changing people 's habits !

490: bugs . when browsing news website , open news in new tab , the Lab will give wired actions , such as recognize the new tab as another website . Seems it can not fit "multi tab of one website" feature .

491: was using a gig of RAM

492: Didn't seem to save and utilize my settings , and did not sync across my multiple browser instances .

493: didn't hide dashboard , comments and suggestions on youtube

494: Worked briefly , but eventually it just became too easy to ignore the interventions .

495: testing

496: Screen doesn't seems to be fit when using !

497: didn't need anymore

498: it was good at first but then i started ignoring it so on to something else unfortunately

499: It simply was not working

500: I realized I didn't spend a lot of time on the websites I was controlling so no need for the extension .

501: not what i was looking for

502: It didn't prevent me from going to webs I didn't want to visit , I visited them anyway and just turned off the extension . It wasn't strong enough

503: I felt that the nudges always changing wasn't affective - mainly got it to stop the feed on youtube from showing , but it kept on popping up again... otherwise nice app !

504: is bugged. can't delete site from the list and can't add it

505: Because use requests to modify sites. It may make the server have a big load of requests, some of them block my access because of that. Like Dribbble

506: wasn't using it

507: Every time I visited a site, it would ask me for the low/medium/high/etc on how it should handle things. That needs to be remembered. Got obnoxious.

508: tg

509: i dont need it

510: I do not know how to use it

511: Found it wasn't necessary.

512: im just lazy and not intersted in improving myself

513: I feel it's an emotional problem. Not a time management problem.  
Thank you.

514: I just wasn't using it. i installed it and didn't do anything with it.

515: didnt want to set it up on this computer. still on my other systems

516: legit bad

517: need firefox version. chrome sucks

518: battery drain

519: It stopped working after a while. Even the most effective of them.

520: Huge RAM usage (250MB) in Opera browser on Mac (macOS High Sierra).

521: Problem solving - will reload when problem is solves.

522: settings are confusing.

523: I was hoping to be able to do something like pomodoro method with this

524: didnt work

525: suggestion to choose level of intervention keeps popping up despite me choosing it

526: sdasda

527: no reason i was just trying it if i feel like actually using it i will reinstall it

528: I wasn't using it

529: at first it was annoying and i noticed a lot of lag time on loading pages which was more annoying than anything, then it was there as a reminder or more like a notification, then it just didn't really help at all

530: I was tired of seeing a countdown on my gmail. I turned off all alerts for gmail, but it was still there.

531: the time counter continues to work even when I'm afk

532: trying to install it again, just did not work

533: The prompt about which mode to use kept reappearing.

534: I have no time do evaluate this plugin.

535: My PC started having Blue Screen failures/crashing.

536: it was good but i give up im lazy :/

537: crashes

538: Disabling because I didn't need to be super focussed anymore, was useful for final exams season though!

539: Takes up way too much memory/RAM.

540: Estoy probando con otra app

541: HabitLab was causing some issues with my browser and seemed to be seriously slowing down my computer. Really liked the extension, which did a great job in preventing distractions, but the computer slow down was too much of a trade off.

542: Just didn't need it. Good job anyway, thx. :)

543: When I am in that downward spiral of distraction, I see the interventions of HabitLab as an externally imposed authority to rebel against. This makes me want to engage in the distracting websites even more.

544: Slowing down my computer.

545: I'm using LinkedIn, twitter for reading professional content. So tracking the way I consult doesn't work for me.  
Thanks

546: I've been using browsers other than Chrome lately, so the information from HabitLab isn't terribly accurate or helpful.

547: It works fine, I just didn't feel the need for it anymore (which is probably a sign that it worked). Thanks for helping me out!

548: The interventions are great for when I'm wasting time, but some of the time I actually do need to get on social networking sites and Amazon for work-related tasks (and I don't want to disable HabitLab for the rest of the day, because those sites \*can\* turn into time wasters after work).

549: wanted a site blocker

550: I don't even know what it is trying to achieve... It asked me multiple times how much restriction I wanted – why does it have to ask more than once? And I set it to be a minimal intervention and it was nagging me even after watching just one 5-minute video... Also, I set it to show many different kinds of UI elements on YouTube, but it only ever showed the one blocking the video player before the video starts and after it ends...

I honestly couldn't find any use to it in its current state, even though the concept is really good and I could really use a well implemented version of this... Now, it's just training me how to skip its one and only video-blocking element without ever giving a thought... It has no meaning at all, sorry.

551: I have a website blocker (Cold Turkey) already. I was looking for something that would issue regular reminders to stay focused. Your utility may do that, but I couldn't immediately see how, so I've uninstalled so I can something targeted at that.

552: Didn't use the product much. Its a great concept though!

553: An absolute nightmare. I just wanted to receive reminder of the time connected. Difficult to understand and set up.

554: Battery impact too high

555: esta en ingles

556: don't neeed it

557: Interventions consistently failed to remember preferences.

558: The browser version didn't offer the same stats the app version did , it looked like two different products.

559: Not addiction left

560: I feel like I restrained myself a bit with the app and it's annoying to have around now. Will re-install if I ever feel like I need it again.

561: The Chrome extension LeechBlock better suited my needs for reducing time spent on certain sites.

562: Youtube videos would not load while the plugin was active , even if no interventions were active

563: I'm just an awful procrastinator , really

564: dhedh

565: vote:8

566: <3

I love you guys , and this app and extension are both incredible , but I 'm not in a place where this is effective for me right now. This is really user friendly and intuitive but I noticed I just kept ignoring all interventions so I 'm going back to square one and sort of resetting all my online stuff , including this . Thanks for creating something incredible and I 'm sure I 'll be back!

567: You guys will no longer be harvesting my data. Bad policies and awful app !

568: Only wanted to track time and deactivate comments and sidebar in Youtube. Got tired of the popups and have no idea how to disable them.

569: Kept turning on settings I kept turning off. I just wanted a time keeper, not all the other interventions

570: I'm doing pretty well managing my online time and now the interruptions aren't helpful for me any longer.

571: on a chromebook

572: Moderation. I tried HabitLab hoping to form new browsing habits and slowly change them for the better over time. What turned me off was that the nagging would take place instantly and aggressively, new installs should default to the lowest settings and learn the habits before intervening. Using negative enforcement from the very second you tread on a site does not motivate you to change, it teaches you that HabitLab is a hardline tool that simply annoys and makes you want to turn it off - the exact opposite behaviour you are after. From my experience, fully blocking websites via my DNS was a much more enjoyable way to stop visiting sites cold-turkey and I still have these in place. Definitely consider a delay (5 mins?) before a nagging feature is used and then ramp up the annoyance factor, HabitLab is simply too abrasive out of the box.

573: Install again after some days.

574: may have caused interference with GMAIL working properly

575: hhh

576: obrigado

577: It was very easy to ignore. I need something that is EXTREMELY heavy-handed.

578: I can see how this could be quite useful to others -- it just wasn't for me

579:

580: this is a competitor used by mani people...

Also the nudges are just ok, but the constant feedback you ask for is annoying

581: this is a competitor used by mani people ...

Also the nudges are just ok, but the constant feedback you ask for is annoying

582: bom mais atrapalha

583: Was not tracking my websites properly. It said I would hit goals when I did not.

584: offering only 60 minutes of youtube per day, no choice for HOURS

585: hi

586: dcfhv12jbkwasdrgjhbkqd

587: It was nice to know that I was making an effort to reduce my wasted time, but in the end my "habits" developed to include silencing the habitlab pushes haha

588: dedede

589: it did not allow me to block activity that is problematic. it focused on social media sites where i don't have a problem getting distracted.

590: Interfered with printing every time -- I had to cancel the Print dialog, dismiss HL, and then return to printing.

591: I am using youtube for education and math learning. it's really annoying for me, nice work guys btw.

592: Didn't need it anymore (not a student!)

593: I dont use it

594: could not sign in to keep my preferences across all devices and each time I went on a tracked website I had to answer what my preferences are.

595: i accidentally downloaded it

596: "How aggressive do you want habit lab to be ?" pop up every window

597: Finished university so it was no longer helpful for my needs

598: to much computer resources

599: I had bugs, whan i was on youtube it always asked me for the entensity of nudges ...

600: Sometimes it shows up, sometimes it does not. Also it does not remember my choices .

601: Kept asking for type of intervention on a custom site , no matter how many times a choice was made as to level desired. Level choice was only remembered until the tab was closed.

602: Trying to reset nudge intensity

603: Keeps asking for level of agression

604: Interventions kept repeating after being turned off

605: sadwfdwqfqf

606: I was using it for productivity in college time , it is now the summer and it keeps annoying me.

607: crashed chrome

608: It was super useful to keep track of time spent on websites I knew I was procrastinating on and it got me through exam season really well! (I'll reinstall it when college starts again)

609: I realized I didn't really need it

610: It was glitching and kept asking me which nudge I wanted on each visit even though I had already set the nudges.

611: there are other extensions that provide all time stats

612: man 10 % CPU for this really?

613: A smaller range of simpler interventions would be more effective .

614: jiji

615: Fun to try , but got a little annoying at times . I might be back !

616: Using 800MB of RAM

617: Extensions seemd to use 4-5% of a CPU in BG, impacting battery life .

618: I prefer to use the app version , because I spend most of my time there than on my office desktop .

619: I was hoping to have a specific habit applied to a specific website . For example , I would love to set Facebook at only 3 minutes or the 1 minute assassin . However, this might not work for something like Youtube , which I often use for work . While I love some of the features , having the random approaches didn 't help me and made it seem to get in the way .

620: Uneeded now

621: yep

622: Your product does not compare to our kawakawa fortnite borgar hense forth u and ur mom have a vast variety of diabetes

623: wish it had a mode only to show how long have been on a particular site = I need to go on FB for work – dont want to answer questions all the time – dont want to feel like and addict when really I would prefer to get away from it – just want to know when time spent gets uneconomic

624: they were intrusive and ugly

625: You don 't need to ask everytime if I need a light touch or a heavy hand .

626: I 've enjoyed using it and liked the approach . I liked the timer , but it constantly popped up "how do you want me to behave (minimal -> invasive)" on visits which was annoying ( including items that I had turned off "permanently" (maybe it was per-site , whereas I expected anything turned off like that to be global (perhaps a choice would be good?)

627: Didn 't let me log out of Facebook .

628: I was hoping to see a weekly graph to compare my progress

629: It consumed 1Gb of RAM!

630: did not use

631: Rallenta il mio PC

632: Shit execution. All this did was make me angry at the ham-fisted way in which it tried to block me. Good idea, rethink your approach .

633: ugly

634: Trying to disable because it threw me off Twitter in the middle of a research tweet. Didn't mean to remove it from Chrome.

635: I dont need this anymore. And in this situation , interventions became annoying .

636: Kept asking me about what intervention i wanted to use every time I visited a site , got annoying

637: e

638: The app says it has permission to change and see all the data for websites I visit . This seems to include usernames and passwords .

639: f schifo al cazzo

640: It really helped me during my undergraduate career when I needed to remain focused when I had the ability to wander , but I have graduated now!

641: Returning computer

642: Didn't ever use it

643: I no longer need to stop getting distracted

644: used a lot of cpu , on laptop , was killing my battery

645: fas

646: Interventions were not strong enough – I wish they happened automatically instead of asking me each time how much I wanted it to intervene

647: Interventions were not working

648: I was just testing

649: Fucking annoying popup on every fucking site I visit

650: Didn't really have a use for it.

651: i realized: when I want or need to work, I work, and when I want to relax , I want to do that. This just got in the way.

652: sdvbcx

653: messed up twitter 1password autofill

654: I sopped working :( I'm trying to reinstall it

655: Worked at home and needed it. I don't work at home anymore.

656: Bad Extension and google emailed me saying data was being stolen!

657: So much irritating when you are actually working on your laptop , need to check your email or make some research for work.

658: Job done

659: changed interventions but it seemed to keep asking me how aggressively to do it. i just wanted to track time!

660: no

661: What is the point or the goal of habitlab? I think is senseless you'll people don't like to be bother with anything? they just want to be left alone.....!!!!

662: I never used it.

663: I'm done with school! I only had this installed to keep me focused while I was doing homework

664: <https://github.com/habitlab/habitlab/issues/638>

665: high CPU use

666: My habit has been broken – thank you.

667: Basically , the features don't work the way they're supposed to . I wanted to use this mostly to help me curb excessively media streaming chain-watching (I've gotten rid of my social media some time back) but the timer doesn't work correctly . As soon as you press full screen the timer stops counting so the majority of the time you spend on the tab isn't counted . Also , ideally , you should be able to categorise types of websites and set limits for the category rather than limits being set for an individual website . Otherwise if the goal is to set a 1hr limit per day of media streaming , as the features are currently configured , you could still end up watching an hour of Netflix , an hour of Stan , an hour of Youtube , etc . in a day and still be meeting your configured target according to this tool .

668: Don't feel I waste time enough on sites such as Youtube and Facebook

669: I just went out of my way to dodge all the interventions . It's a ME problem , not an app problem , sorry

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