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InspectorWidget: a System to Analyze Users Behaviors in Their Applications

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Abstract
We propose InspectorWidget, an opensource application to track and analyze users’ behaviors in interactive software. The key contributions of our application are: 1) it works with closed applications that do not provide source code nor scripting capabilities; 2) it covers the whole pipeline of software analysis from logging input events to visual statistics through browsing and programmable annotation; 3) it allows post-recording logging; and 4) it does not require programming skills. To achieve this, InspectorWidget combines low-level event logging (e.g., mouse and keyboard events) and high-level screen features (e.g., interface widgets) captured through computer vision techniques. InspectorWidget benefits end users, usability experts and HCI researchers.

Author Keywords
Logging; automatic annotation; computer vision

ACM Classification Keywords
H.5.2. [Information Interfaces and Presentation (e.g. HCI)]: User Interfaces—Benchmarking

Introduction
Interface designers, HCI researchers or usability experts often need to collect information regarding usage of applications in order to interpret quantitative and behavioral aspects from users or to provide user interface guidelines.
Unfortunately, most existing applications are closed: source code or scripting support are not always available to collect and analyze users’ behaviors in real world scenarios.

Different methods can alleviate this problem such as replicating the application with specific interaction “hooks” [15]. However, implementation is time consuming and the replicated application is often a light version of the real one biasing users’ behaviors. Dedicated activity plugins or ad-hoc accessibility monitors can also capture user events at different levels of granularity [5] but several applications do not support these mechanisms. Moreover, these methods are application dependent, i.e. require a specific programming language and/or implement one plugin for each application which is not ideal for complex task involving several applications. Conversely, screen recording while users carry their task [7] provides a lot of information, but requires to carefully watch and manually annotate the video, which is tedious and not precise enough for studying low-level interactive phenomena (e.g., duration of scrolling actions).

We propose the InspectorWidget, an opensource application to track and analyze users’ behaviors in any closed application by combining computer vision based analysis of video-recording with low-level interaction collection.

InspectorWidget has four key features. 1) It works with all applications, in particular closed applications where the source code is not available, or applications that do not offer plugins support. 2) It supports logs and event browsing offering data analysis at different levels of granularity, from low level (mouse, keyboard events) to high level (screen recording). 3) It supports additional post-logging and analysis because it collects all possible events a priori (screen record, mouse and key events) and that the experimenter defines rules a posteriori depending of what the focus of the study is. Thus, the experimenter can run other studies with different focus using the same collected data, weeks or months later, simply by defining new rules and without having to collect data again. 4) InspectorWidget relies on visual programming [14] and thus minimizes the need of programming skills, making it possible for experimenters with sociological or psychological backgrounds to use it.

To achieve this, InspectorWidget relies on a five-step workflow (figure 1): first, it records users’ screen content and low-level user interactions (mouse and key events) while they carry their task. Second, the experimenter browses this recording and programs annotations on it by combining explicit active zones definition (e.g., “this is the scroller”, “there is the scrollbar”, etc.) and specifying rules (e.g., “the scrollbar is clicked”). InspectorWidget then automates annotations, allowing to combine the programmed ones with computer-vision and low-level events analysis. Finally, the experimenter analyzes the annotations either by using the visualization provided in InspectorWidget, or by exporting data to process it in her favorite statistical analysis tools.

Design Goals: Observing Applications’ Usage
To better understand users’ need and practices, we conducted an informative study with 4 experts. We interviewed two HCI researchers, one usability consultant and one software developer to cover different class of users. Based on these interviews, we present five design goals and discuss related works according to these design goals (tables 1 and 2 summarize these goals and compare existing systems).

Collect diverse data. Understanding subtle interactions generally requires to study both users’ actions (e.g. mouse, key events) and the semantics of the interface (e.g. menu, buttons). It is thus necessary to collect diverse data such as:

- keylogging or macro recording, for low-level interactions such as mouse or keyboard events;
Monitor diverse applications
Collect diverse data
Automate diverse annotations
Browse collected data
Allow iterative monitoring
Minimize programming requirements
Platforms
Distribution

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<th>Collect diverse data</th>
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Table 1: Comparison between InspectorWidget and related works. Gray items indicate partial implementations or plans for future work.

- widgets and states changes logging, for applications implementing accessibility Application Programming Interfaces (APIs) that are specific to operating systems; or applications providing their own information (e.g. Document Object Model (DOM)).
- screencasting, to get a video record of the interface.

AppMonitor [5] combines the two first methods to monitor Microsoft Word and Adobe Reader. However, it requires two different implementations as the "accessibility monitor" differs between these two applications. The Autodesk Screencast based on the Chronicle project [10] is a powerful tool combining the three methods but it is limited to Autodesk applications only. Screen capture software such as ScreenFlow record both video and key events, events can be overlaid on the video but not exported as raw data.

Browse collected data. Browsing collected data is an important part of data analysis. With Chronicle [10], users can visualize all event histories on an interactive timeline and play back the actual workflow. Delta [12] allows to compare workflows through several views, among which a cluster of similar workflows and a union graph of shared commands. Finally, Patina [13] overlays a heatmap of mouse locations directly over the recorded GUI.

Automate diverse annotations. Combining screen recording with computer vision can be used to reverse engineer the GUI of closed applications. This method has been proven useful for automating GUIs and their testing [8], for better understanding the GUI [6], or for regenerating GUIs [9].

Capitalize through a posteriori annotations. During data analysis, experimenters can think of different study that is unfortunately impossible to conduct and would require to re-collect data. Typically, a usability expert could investigate how users display their slides when editing them in PowerPoint, and then realize that she was interested in how users include external materials in their slides. Efficient application usage observation should capitalize on collected data. In that respect, a tool recording all data such as the screen and the entire DOM description of the interface, where the experimenter specifies the focus of the study a posteriori, is more powerful and flexible than a dedicated logging plugin (e.g. to study a scrollbar position) or a tool where the experimenter specifies a priori what she wants to collect.
Minimize programming requirements. Experimenters, analysts or usability experts are often not computer scientists. They are thus not always familiar with programming language. Even computer scientists are not familiar with all programming languages. A software observing usage application should minimize the needs of programming skills to favor immediate usability and wide adoption.

InspectorWidget
In this section, we describe the concept and implementation of our application.

Concept
Analyzing user’s behaviors with closed applications using InspectorWidget follows a five-step workflow: 1) record activity, 2) browse and 3) program annotations, 4) automate annotations and 5) analyze the results (figure 1).

Record activity. First, the experimenter records the screen and logs events to track users activity. Screen recording and interaction events logging are usually separated into two tools, making the observation process cumbersome and the collected data harder to synchronize. We solve this problem by proposing a cross-platform desktop tool that both records the content displayed on screen(s) and logs low-level interaction events (mouse and keyboard).

Browse and Program annotations. The experimenter then programs annotations. In practice, she browses a recording to find a time frame where the system was in a specific state she is interested in (figure 2). She then draws bounding boxes on widgets of the recording for specifying a center of interest, and creates rules on this object with a dedicated syntax (figure 3). Using our example scenario, the experimenter wants to know which color selection tool users used. She first browses the recording until the color picker window of Powerpoint is displayed; she then draws bounding boxes around all of the five color tool buttons (A* on figure 2) for each selection state (ColorWheel, sliders, spectrum, etc.). Because she is only interested in when users were interacting with a color tool, she wants InspectorWidget to annotate only when the mouse pointer was located over the color tool window. In that purpose, she also annotates the color dropper (B), the window buttons (C) and the window title (D). These graphical widgets are always displayed in the color tool window (E) and can be tracked using visual programming (figure 3) to retrieve the size and location of the window even if it has been moved or resized.

Figure 2: Programming annotations (concept): the experimenter defines an active region for the five color tools selection (A*), and then defines active regions for the color dropper (B), the window buttons (C) and the window title (D) that are used to retrieve the size and location of the window even if it has been moved or resized.
Figure 3: Programming annotations (implementation): on the top the experimenter creates rules using a visual programming language to program the automated annotations, on the bottom the annotated timeline reveals from 1:22 that the user toggled colors tools from sliders to palette then spectrum but not disc.

Note that browsing the video to locate the time where the interface is in a specific state (e.g. the color picker is displayed) might be a difficult task because the experimenter does not know exactly when the interface was in this state, or because it is possible that the interface was never in this state. For this reason, the experimenter sets the interface in the correct states on her own computer and then records it. She can also program the annotations and use these as an input for InspectorWidget to annotate all users’ recordings.

Automate annotations. Once all annotations have been programmed, the experimenter clicks on "automatic annotation" to instruct InspectorWidget to start the automatic annotation process, using computer vision and the recorded low-level and programmed annotations as inputs (see below for implementation details).

Analyze. The experimenter can directly visualize simple statistics before exporting results. InspectorWidget produces interactive plots on several extracted and nested metrics, such as the time the mouse pointer was over a widget. Data can also be exported as csv files to be processed in the experimenter’s favorite statistical analysis tool.

Implementation
InspectorWidget is composed of a cross-platform desktop application for recording, a dedicated server for automating annotations, and a web-based application for browsing the recordings, programming annotations and analyzing the results. InspectorWidget is released as an opensource project under a GPLv3 license, available on github.


Browse recordings, program annotations. Browsing recordings and programming annotations is achieved through a dedicated web-based application, based on the amalia javascript video player [11] for browsing and enriching the video with annotation, and based on Google Blockly (inspired by Scratch [14]) for programming the annotations.

Automate annotations. The automatic annotation of the recording is performed by a dedicated server, using OpenCV for template matching and tesseract [4] for text/number detection.

Analyze. The aforementioned web-based application also provides simple data visualization by displaying charts rendered using the d3.js javascript library.

1http://github.com/InspectorWidget/InspectorWidget
Discussion and future work
We see several directions to improve these works including technical considerations, novel features and user studies.

Technical considerations. A limitation of our approach is that it relies on computer vision which can require time to process large screen videos. We plan to speed up this computational phase with GPU/Parallel processing since we can process the video frames in a non-linear sequence and after the recording. We also plan to pre-process the video in order to remove sequences without users’ actions. This will improve both human processing (less video to watch) and computer vision processing. Finally, we plan to delegate some video-based analysis (e.g. position of a pop-up menu) to accessibility tools when these are available on the target application / operating system. Indeed, extracting DOM data can provide information about the hierarchical structure of the widgets and accelerate the computer vision process.

Features. Extracting DOM information will also provide an additional level of granularity between low-level events (e.g. mouse and key events) and high-level events (video screen) by considering the attributes of graphical widgets. We also plan to extend our software from desktop workstation to mobile devices.

Evaluation (of annotated data). As every computing system relying on computer vision, InspectorWidget is likely to result in errors, either by missing parts of the videos that should be annotated, or by incorrectly annotating parts that should not. We thus plan to run an evaluation with various use cases in order to quantify the accuracy and precision of InspectorWidget, assessing its efficacy in terms of usability.

Evaluation (through annotation tasks). We also plan to conduct a user study with HCI researchers and usability experts to evaluate the qualitative and quantitative benefits of InspectorWidget over manual annotation, assessing the efficiency of our system and the satisfaction of its users. One of the primary motivations for creating InspectorWidget was to support longitudinal “in the wild” studies concerning our own research projects. Reciprocally, the analysis of such upcoming studies will provide insight on the strengths and shortcomings of InspectorWidget.

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