X-PERT: A Web Application Testing Tool for Cross-Browser Inconsistency Detection*

Shauvik Roy Choudhary†, Mukul R. Prasad§, Alessandro Orso†

†College of Computing
Georgia Institute of Technology
Atlanta, GA, USA
{shauvik | orso}@cc.gatech.edu

§Software Systems Innovation Group
Fujitsu Laboratories of America
Sunnyvale, CA, USA
mukul@us.fujitsu.com

ABSTRACT

Web applications are popular among developers because of their ease of development and deployment through the ubiquitous web browsing platforms. However, differences in a web application's execution across different web browsers can cause cross-browser inconsistencies (XBIs), which are a serious concern for web developers. Identifying XBIs manually is a laborious and error-prone process. In this demo we present X-PERT a tool for identifying XBIs in web applications automatically, without requiring any effort from the developer. X-PERT implements a comprehensive technique for identifying XBIs and has been shown to be effective in detecting real-world XBIs in our empirical evaluation. The source code of X-PERT and XBI reports from our evaluation are available at `http://gatech.github.io/xpert`.

Categories and Subject Descriptors: D.2.5 [Software Engineering]: Testing and Debugging

General Terms: Reliability, Verification

Keywords: Web Testing, Cross-browser Testing, Layout Testing

1. INTRODUCTION

Web applications are increasingly being used for both personal and business activities. Users of such applications might use any web browser to access them, and the application is expected to behave consistently across these different environments. However, web applications often exhibit differences when executed in different browsers, leading to cross-browser inconsistencies (XBIs). XBIs are discrepancies between a web application's appearance, behavior, or both, when it is run on two different environments. XBIs are not only fairly common, but also notoriously difficult to identify and fix. For example, 5328 posts were created and tagged with "cross-browser", on `stackoverflow.com` over the past four years alone. Moreover, nearly 2000 of these posts have been active during the past year [8].

In general, if XBIs are not identified during testing, they can adversely degrade the experience of the users of the web application with the affected browser. In fact, as shown in our evaluation of X-PERT, some XBIs completely prevent users from accessing the functionality offered by the web application, thereby rendering it useless on that particular platform. XBIs are thus a serious concern for companies, which rely on such applications for business or for creating their public brand image. The current practice in industry is to identify XBIs through manual inspection of the web application screens across all the different browsers [2]. Such testing is not only human intensive, but also error-prone.

Recent work on identifying XBIs [7, 3, 5] has proposed techniques that go beyond the state of the practice, focus only on certain aspects of a web application's execution, and are well suited for specific types of XBIs. For instance, the WebDiff tool [7] uses computer vision to detect XBIs, whereas CrossT [3] uses graph isomorphism along with text comparison to find XBIs. These tools, however, offer only partial and imprecise solutions to the XBI detection problem. To address these limitations of existing techniques, we proposed a technique, implemented in X-PERT, that integrates a rich set of comparison techniques and orchestrates them to apply each technique to the class of XBIs that it is best suited to detect [6]. Our technique is an automated, precise, and comprehensive approach for XBI detection and is based on our findings from an extensive study of XBIs in real-world applications.

This demo paper discusses the architecture and implementation details of the X-PERT tool and is organized as follows. Section 2 describes the different kinds of XBIs. Next, we summarize the technique implemented in X-PERT in Section 3. Section 4 presents details of our tool's implementation along with its usage scenario. The evaluation of X-PERT and related work are presented in Sections 5 and 6. Finally, we conclude in Section 7.

2. CROSS-BROWSER INCONSISTENCIES

To get a deeper understanding of XBIs, we performed a systematic study of 100 real-world web applications [6]. Through this study, we were able to establish a classification of XBIs, which further helped us in defining our technique, described in the next section. In particular, we found three main types of XBIs: structural, content, and behavior.

*This demo illustrates the implementation of a technique presented at ICSE'13 [6].

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Structural XBIs: Such XBIs affect the structure, or layout, of individual web pages. The web page structure is essentially a particular arrangement of elements, which in case of structural XBIs is erroneous in a particular browser. For instance, the misalignment of one or more web page elements on a given web page, in a particular browser, can constitute a structural XBI. We found that this was the most common category of XBIs, occurring in 57% of the subjects with XBIs.

Content XBIs: This kind of XBI is observed in the content of individual components on a web page. Such differences can occur, where the visual appearance of a web page element, or the textual value of an element, are different across two browsers. We further classify these two cases as *visual-content* and *text-content* XBIs. In our study, we found that these XBIs occurred in 30% and 22% of the sites with XBIs respectively.

Behavioral XBIs: These type of XBIs involve differences in the behavior of individual widgets on a web page. An example of such an XBI would be a button that performs a particular action within one browser and a totally different action, or no action at all, in another browser. Another example of behavioral XBI is the presence of an HTML link, which works in one browser but is broken in another one. In our study, such XBIs occurred in 9% of the web applications with XBIs.

In summary, behavioral XBIs affect the functionality of individual components, resulting in broken navigation between different screens. Structural and content XBIs, conversely, involve differences in the arrangement or rendering of elements on a particular web page. In the next section, we describe how our technique detects each of these XBIs.

3. TECHNIQUE OVERVIEW

Algorithm 1 presents an overview of our XBI detection technique. As shown in the algorithm, our technique takes as input the URL of the home page of the web application under test, *url*, and two browsers considered for the testing, Br_1 and Br_2. The technique outputs a list of XBIs, $\mathcal{X}$. In this paper, we only summarize the main steps of the algorithm (for all the details, see Reference [6]).

Model Generation via Crawling: The technique starts by crawling the web application, in an identical fashion, in each of the two browsers Br_1 and Br_2. In this process, it records the observed behavior as navigation models M_1 and M_2. The model is captured as a labeled transition system, which represents the top-level structure of the crawled web application. In the model, the states correspond to web application screens, and each transition is labeled with a widget action that leads to a screen navigation. In addition to this navigation model, we also capture the screen image and the DOM structure of the elements on each observed screen. In the algorithm, this step is implemented in function *genCrawlModel* (line 3).

Behavioral XBI Detection: The navigation models M_1 and M_2 are checked for equivalence to uncover differences in behavior. To do this, the technique uses the graph isomorphism checking algorithm for rooted labeled directed graphs proposed in [3]. This algorithm is implemented in the *diffStateGraphs* function (line 4), which produces a set of differences ($\mathcal{B}$) and a list *PageMatchList* of corresponding web-page pairs S_i^1, S_i^2 between M_1 and M_2. $\mathcal{B}$ contains a set of missing and/or mismatched transitions across pages,

Algorithm 1: X-PERT: Overall algorithm

```
Input  : url: URL of target web application
         Br1, Br2: Two browsers
Output: X: List of XBIs
1  begin
2      X ← ∅
3      (M1, M2) ← genCrawlModel(url, Br1, Br2)
       // Compare State Graphs
4      (B, PageMatchList) ← diffStateGraphs(M1, M2)
5      addErrors(B, X)
6      foreach (S_i^1, S_i^2) ∈ PageMatchList do
           // Compare matched web-page pair
7          DomMatchList_i ← matchDOMs(S_i^1, S_i^2)
8          L_i^R ← diffRelativeLayouts(S_i^1, S_i^2, DomMatchList_i)
9          C_i^T ← diffTextContent(S_i^1, S_i^2, DomMatchList_i)
10         C_i^V ← diffVisualContent(S_i^1, S_i^2, DomMatchList_i)
11         addErrors(L_i^R, C_i^V, C_i^T, X);
12     return X
```

representing differences in dynamic behavior. Thus, $\mathcal{B}$ represents the behavioral XBIs detected by the algorithm and is included in the final output $\mathcal{X}$. *PageMatchList* contains the mapping between corresponding screens across browsers and is used to detect other kinds of XBIs (lines $6-13$).

Matching screen elements: To be able to find XBIs on two matched pages S_i^1 and S_i^2, X-PERT first computes a list of corresponding DOM element pairs in these pages ($DomMatchList_i$). This computation is performed by function *matchDOMs* (line 7) and is based on a *match index* metric for DOM element correspondence. The same metric, which has been used in earlier work on XBIs [7, 5], is a value in the range $[0, 1]$ and is computed using a weighted combination of (1) the element's DOM attributes, (2) its XPath (*i.e.,* path in the DOM tree—`http://www.w3.org/TR/xpath/`), and (3) a hash value computed from its descendants in the DOM tree. (See Reference [5] for further details.)

Structural XBI Detection: A key contribution of the X-PERT technique was the notion of an alignment graph, which is an abstraction of the layout of a web page that captures the relative arrangement of all the elements on that page. X-PERT extracts alignment graphs from different renderings of a given web page on different browsers and compares them to detect structural XBIs. This is implemented by *diffRelativeLayouts* (line 8 in Algorithm 1), which compares pages S_i^1 and S_i^2 and extracts the set of relative-layout differences $\mathcal{L}_i^R$ that represent structural XBIs. (More details about this technique can be found in Reference [6].)

Text-content XBI Detection: This step detects textual discrepancies in web page elements that contain text. To detect this class of XBIs, the text-value of an element is extracted from its DOM representation and compared with that of its corresponding element from $DomMatchList_i$. In the algorithm, *diffTextContent* (line 9) implements this checking and is computed over all the text bearing leaf nodes in the DOM tree. (For more details, see the *LDTD* feature for machine learning in [5].)

Visual-content XBI Detection: These XBIs represent differences in the visual appearance of individual page elements (*e.g.,* differences in the styling of text or background of an element across different browsers correspond to visual-content XBIs). Such errors can only be observed in the image representation of the elements. Hence, the technique measures the χ^2 distance between the color histograms of the element's screen image. We also used this approach in

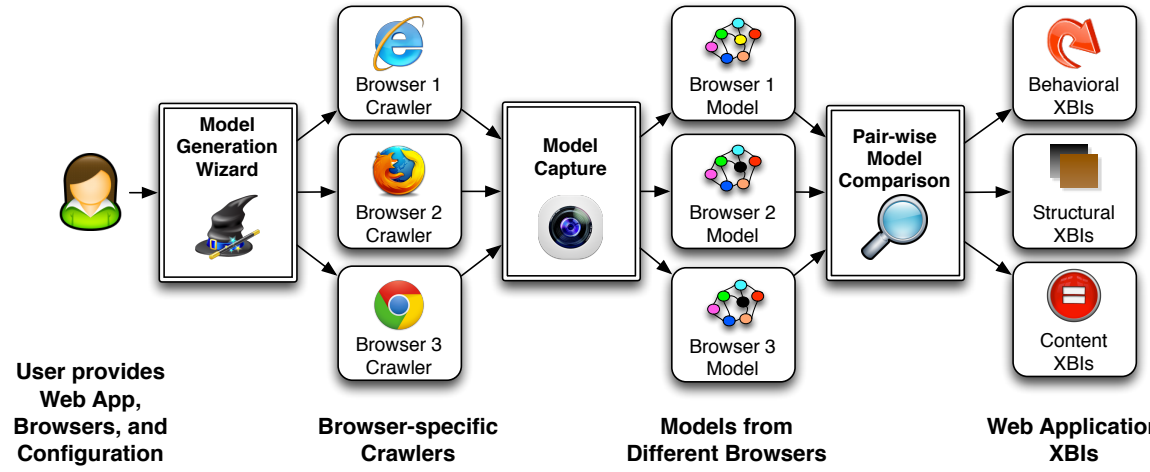

Figure 1: High-level overview of X-PERT.

our CrossCheck approach [5]. However, in X-PERT, we only apply this to the leaf DOM elements, where it is most effective at detecting visual-content XBIs. This operation is implemented by function *diffVisualContent* (line 10).

All the XBIs detected by the technique are added to the XBI list $\mathcal{X}$ (line 11) and reported to the developers.

4. TOOL DESCRIPTION

X-PERT can work with any web application that runs on desktop browsers. Since X-PERT analyzes the client-side of such applications, it is agnostic to any server-side technology. X-PERT is written in Python and Java and can run on a variety of desktop operating systems, including Windows, Mac OS X, and Linux.

Figure 1 shows a high-level overview of X-PERT, which operates as follows. First, the user invokes the web interface of the tool and interacts with its model generation wizard. This web interface is implemented in Python using the Flask framework (`http://flask.pocoo.org`) on the server-side, and Twitter bootstrap (`http://getbootstrap.com`) and jQuery (`http://jquery.com`) libraries on the client-side.

Once the user submits the subject web application's URL and model capture parameters to the wizard, X-PERT uses this information to generate different crawler instances, one for each browser. The generated models are then processed by the model comparison module, which applies our proposed technique to compare these models in a pair-wise fashion. This model comparison module is a key contribution of the X-PERT technique as it compares the different aspects of the web application's execution to uncover the three types of XBIs, which are then gathered, tabulated, and reported to the user.

The architecture of X-PERT, shown in Figure 2, consists of the model capture and comparison modules. Both these modules are mainly implemented in Java. Further details of the implementation are discussed below.

- The *Model Capture* module uses the CrawlJax tool [4], which internally uses the Selenium testing framework (`http://seleniumhq.org`) to explore the web application in the different web browsers. We extended CrawlJax to save the model from its exploration along with the *screenshot* and *DOM structure* of each page. The DOM structure is obtained by querying the browser through its JavaScript interface and contains properties of each web page element's rendition. These properties include, for each element, its textual content, style information, its XPath location, its size, and its coordinates on screen.

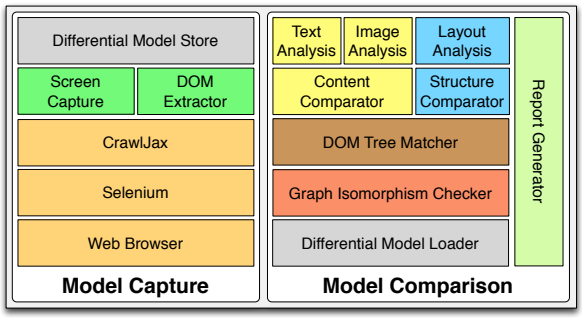

Figure 2: Architecture of X-PERT.

- The *Differential Model Store / Loader* components are used to persist and load the models to/from the file system as XML files. These components are implemented using the Object serialization support in Java and are essential for the model capture and comparison components to operate independently. For instance, model capture might be run to collect models from different machines, whereas model comparison may compare these models on a single machine.
- *Graph Isomorphism Checker:* This module performs an equivalence check between the graph based models from the two browsers to identify mismatched states and transitions across the two models. These models are implemented as Java objects and compared through a pair-wise traversal of the two graphs.
- The *DOM Tree Matcher* module matches corresponding elements on renderings of a web page across the different browsers by computing the *match index* metric. This metric considers the Levenshtein distance between the XPaths of the elements, and is computed using the corresponding implementation in the Apache StringUtils library.
- *Content Comparator:* This module performs textual analysis of corresponding elements to detect text-content XBIs. For detecting image-content XBIs, it compares screen images of the corresponding elements on the web page by leveraging the OpenCV toolkit [1]. Specifically,

Table 1: Results of X-PERT's empirical evaluation.

Name	Behav.		Struct.		Content				Total	
					Text		Image			
	T	F	T	F	T	F	T	F	T	F
Organizer	1	0	9	0	0	0	0	0	10	0
GrantaBooks	16	0	11	0	0	0	0	0	27	0
DesignTrust	2	0	5	3	0	0	0	0	7	3
DivineLife	7	0	3	6	1	0	0	0	11	6
SaiBaba	2	0	2	9	0	0	0	0	4	9
Breakaway	0	0	10	2	0	0	0	0	10	2
Conference	2	0	3	0	1	0	1	0	7	0
Fisherman	1	0	3	1	0	1	1	0	5	2
Valleyforge	0	0	2	2	0	0	1	0	3	2
UniMelb	2	0	0	0	0	0	0	1	2	1
Konqueror	0	0	0	0	0	0	0	6	0	6
UBC	0	0	0	0	0	0	0	0	0	0
BMVBS	0	0	0	0	0	0	0	0	0	0
StarWars	0	0	12	0	0	0	0	0	12	0
TOTAL	33	0	60	23	2	1	3	7	98	31

this image comparison measures the χ^2 distance between their color histograms to detect image-content XBIs.

- *Structure Comparator:* The structure of the page extracted by the model capture module is analyzed by the Layout Analysis module to create alignment graphs, which represent the relative alignment of web page elements. Another graph isomorphism checker was implemented in Java to find differences in the alignment graphs of corresponding screens.
- *Report Generator:* This module generates an HTML report tabulating the set of detected XBIs. It is implemented using the Apache Velocity library (`http://velocity.apache.org`). The generated reports leverage jQuery along with the HTML5 Canvas for rendering and highlighting the XBIs on the screen images. Behavioral XBIs are presented in the HTML report by highlighting them on the models converted to SVG format using the GraphViz tool (`http://www.graphviz.org/`).

5. EVALUATION

To assess the usefulness of X-PERT, we ran it on 14 subjects. These subjects are divided in three groups: the first six subjects were used in prior work, the next four were from our study, and the final four were obtained using an online random URL service (`http://www.uroulette.com/`).

Our experiments were performed using the latest stable versions of Internet Explorer (v9.0.9) and Mozilla Firefox (v14.0.1). The results of our investigation of X-PERT's effectiveness are shown in Table 1, which lists, for each subject, the XBIs reported in the terms of true (T) and false (F) positives. As shown in the table, X-PERT was effective in finding different kinds of XBIs in the subjects. A deeper investigation of the results [6] revealed that X-PERT's precision and recall are 76% and 95%, respectively, against 18% and 83% for the state-of-the-art tool CrossCheck [5].

6. RELATED WORK

To the best of our knowledge, X-PERT is the first tool for comprehensive detection of XBIs. Previous research tools (*e.g.,* [7, 5, 3]) only focused on certain types of issues and had low precision and recall. Developers typically use browser-compatibility tables, such as `Quirksmode.org` and `CanIUse.com`, to check their web applications. Some web development tools, such as Adobe Dreamweaver (`http://adobe.com/products/dreamweaver.html`), provide basic static-analysis based hints to help detect certain issues. However, the issues targeted by reference websites and development tools are limited to features that are known to be missing in a particular browser. Other tools, such as BrowserShots (`BrowserShots.org`) and Microsoft Expression Web SuperPreview (`http://microsoft.com`), provide previews of single pages in different browsers, while tools such as `CrossBrowserTesting.com` and `BrowserStack.com` allow for browsing web applications in different emulated environments. In both cases, the comparison of the observed behavior across browsers must still be performed manually.

7. CONCLUSION

Cross-browser inconsistencies (XBIs) are a serious problem for web developers. Current industrial practice relies on (expensive and error prone) manual inspection to find these issues. Existing research tools, conversely, only target particular aspects of XBIs and can report a significant number of false positives and negatives. To address these limitations, we presented X-PERT, an open source tool for comprehensive XBI detection. Our empirical evaluation shows the effectiveness of X-PERT over the state of the art. This demonstration presents the details of the implementation of X-PERT and illustrates how it is fully automated and easy to use through its web interface. In addition, X-PERT generates easy to comprehend and actionable reports for the developer, thus allowing them to address XBIs effectively.

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