

Detecting Features Inconsistency in Cross-Platform Applications

C.Shanthi, K.Sharmila, J.Jebathangam, R.Devi

Abstract: The development of mobile computing has offered rise to a differing set of computing platforms. Customers utilize these distinctive stages for both personal and business exercises, for example, banking, shopping and so on. The developers of mobile computing faces many challenges which includes numerous versions of operating systems and thousands of devices which various in screen sizes. While developing mobile applications, the application engineers have to determine APIs of a home stage (e.g., Windows Phone), and consequently create forms of the application for different target stages (e.g., iOS and Android). Because of this cross platform application development is a striking issue for programming engineers who need to pitch to clients regardless of which platform they run (Windows, Mac or Linux).

Keywords: Mobile Application, Cross Platform, Inconsistency, Android, iOS

I. INTRODUCTION

Maximum number of mobile devices is operated on Android, iOS or Windows 10; these are generally known as operating systems or platforms. The three types of mobile application namely native application, web application and hybrid application. Native applications are created to aim one particular platform like android, iOS or Windows. Mobile web applications are developed for web applications to deliver pages on web browsers which are getting operated in mobile devices. Hybrid applications have been developed to aim multiple Cross platforms mobile applications. The major dispute in developing Hybrid/cross platform mobile applications for interoperability through multiple platforms is to retain the developed application consistent across different platforms. The next frequent issue in a cross-platform mobile application is to identify the missing features among the applications which are developed for different platforms. The application developers should check the developed mobile application on each and every platform independently and physically executing screen-by-screen assessment and detecting many inconsistencies in cross-platforms. But it is very difficult, time consuming and moreover it might have errors.

II. LITRETURE SURVEY

The problem of solving multiple user interfaces started a while

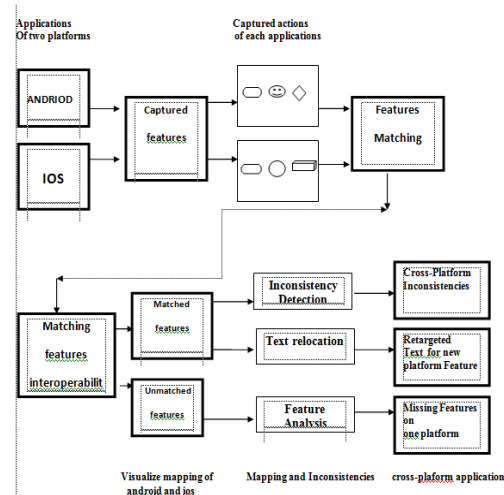


Fig. 1. Overview of the Technique

ago, however, to date actual insufficient effort is available for mobile devices explicitly in the literature. Mattia Fazzini and Alessandro Orso (2017) suggested DIFFDROID, an innovative method that assistances developers to discover inconsistencies in native mobile applications. DIFFDROID syndicates contribution generation and discrepancy testing to assess the performance of an application on different platforms and identifies possible inconsistencies Choudhary et al. (2014) developed a technique to examine the client-server communication and network traces of different versions of a web application to match features across platforms. A.Mesbah and M. R. Prasad (2011) found an automated solution for the problem of cross-browser compatibility testing of modern web applications as a 'functional consistency' check of web application behavior across different web browsers. Their method involves mechanically examining the assumed web application under dissimilar browser surroundings and taking the behavior as a finite-state machine and formally comparing the generated models for equivalence on a pair wise-basis and exposing any observed discrepancies. L. Wei, Y. Liu, and S.C. Cheung (2016) have reported a technique named FicFinder to detect compatibility issues in Android applications. .

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III. DETECTING FEATURES INCONSISTENCY

This technique recognizes and coordinates the highlights of a cross-platform mobile application by examining the customer server correspondence that happens when the mobile application is utilized on the distinctive platforms. At a higher level, the technique operates in four major steps: (1) to verify the traces of the system contact among the client and server of various platform-specific versions of a cross platform mobile applications, (2) to find each traces and unique states in the model as a sequence, (3) to recognize a division of these traces as feature instantiations, (4) to match the feature sets recognized for each and every platform-specific version of the cross platform mobile application which is used to recognize 78 matched and lost features, and different functionality of applications across versions, (5) to create a image of the different models, finding the detected data inconsistencies. The Fig 1 shows the overview of the featCHECK. It captures the activities of the Android and iOS application running on two unique stages and various Shapes have been captured for the two platforms discretely. It can be an obviously expressed model or can be the runtime trace of the mobile application. At this point, reasonable captured data is then compared over numerous platforms and can prompt both matched and unmatched behavior over the platforms. Which matches the number of edges and unique states in the shapes of the mapping features.

A. featCHECK

The featCHECK is used for exactly identifying matched and un-matched features finding in cross platform mobile application. Figure 1 gives a high-level vision of the technique used, featCHECK. The first and foremost step of featCHECK is to group a set of system level traces for both mobile application versions. The resulting feature mapping has been developed by the premise of trace-sets. The fundamental process of mapping features generally independent and based on trace collection. There are three principal stages, reflecting the three difficulties discussed which has been given below. The primary step, the system traces are found to recognize request which are examples of the same action. In this stage, all requests are preoccupied and mapped onto a little alphabet of actions. In the next step, the unique traces from each and every platform are bundled and canonicalized into a center arrangement of traces. In the third step, the canonicalized traces from the cross platform mobile applications are looked at against each other to locate mapping between features.

B. Algorithm: 1

```

I/p: T: Locate traces
O/p: CA: Collection of actions
Begin
    K ← TraceSimplify(T)
    Level 1 Collection
    Level 1Collection ← SimpleCollection(T; url-path equals)
    Level 2 Collection
    Level 2 Collection ← {}
    TD ← {TanimotoDistance(k1; k2) | k1; k2 ∈ K}
    underCollection ← split(Level 1Collection; size = 1)

```

```

overCollection ← split(Level 1Collection; size > 1)
L2Collection:
add(AgглоCollection(underCollection; TD; (<; t1)))
For each c ∈ overCluster do
    L2Collection: add(AgглоCollection(c; TD; (>; t2))
return L2Collection

```

C. Algorithm: 2

```

Trace Simplify
I/p : T: Set of cross platform traces = {T1, T2, ..., Tn}
O/p: W: Set of words tuple sequences = {w1, w2, ..., wm}
Begin
    W ← ( )
    For each T ∈ T do
        For each <request, reaction> ∈ T do
            while isRedirect(reaction.code) do
                reaction ← followRedirect(reaction)
                if isCodeOrData(reaction.type) then
                    w ← getWordsw(request.path, request.qs)
                    W.add(w)
Return w.

```

In iOS and Android have diverse User Interface (UI) components; mapping process is expected to discover identical gadgets. The Graphical User Interface (GUI) components that exist for both local Android and iPhone platforms are distinguished based on the dissimilarity and similarities on both the platforms. These UI identical mappings are made freely accessible. The relation returns 1 if both components are viewed as mapped and 0 if not. The traces and feature mapping stages which yields individual models with an arrangement of registered secondary mapping properties for their unique states and limits (FMIG and FMAG). This algorithm works based on the following assumptions (1) the application model begins with first edge that prompts an initial state and (2) competently, the two models begin with a similar starting states and edge Pairs of iPhone and Android. In order to discover the edge-pairs, all the outgoing iPhone edges are acquired. The feature mapping illustrates the matching problem which has the extreme weighted bipartite coordinating (MWB) issue. It gives a bipartite chart $G = (V: E)$ where V and E represents Vertices and Edges respectively. The information for states and edges are utilized as a part of this phase to outline feature of two models. The Fig 2 demonstrates unmatched features from both the platforms

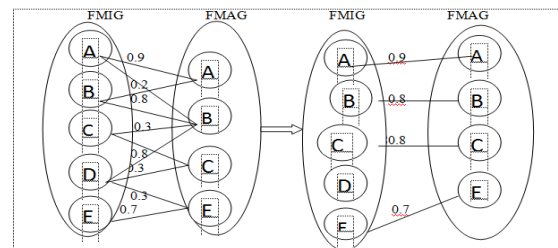


Fig.2. Bipartite graphs of features mapping

This issue is to find out similar of maximum weight wherever the weight of matching M is given by $w(M) = Pe^{2M} w(e)$.

The definition of the MWBM is utilized in Hungarian Algorithm. The edge E operates amongst iOS and ANDROID and signifies the opportunity of similar and comparing features. The weight on an edge denotes the profits of matching both the features, the possibility is that they are surely right matches. On the other side, features A-E from the iOS Platform (FMIG) are connected with features 1-4 from the Android platform (FMAG) through edges, labeled for each pair. On the right side of the figure 3 is the answer for the MWBM problem where only the edges adding to the extreme profit are engaged. This process of matching is the last and final consequence of the calculation and gives a summary of matched features, which is [(A, B, C, D, E); (A, B, C, E)] for the representation.. Table I shows the collective numeral of 'Unique States', 'Edges', User Interface essentials aimed at all the circumstances operating on every Android and iPhone application, developed by featCHECK. The last splinter of the table exhibits the quantity of states which are unique and the result demonstrates that featCHECK can differentiate the new constraints of a specified iPhone and Android application-pair and creates a couples of situations both in iPhone applications and Android applications wherever the quantity of physical noteworthy exclusive states which do not accurately synchronize the quantity of states which are exceptional and poised through the vibrant analyzer. This is primarily approach now receipts into account the type of the class (both chat in Android and missing chat in iOS) in major a unique state and thus divided for different views.

The Table above makes to evaluate a step by step process, in finding the traces of the applications which makes the task easy to find the data inconsistencies in given 25 open-sources cross platform applications.

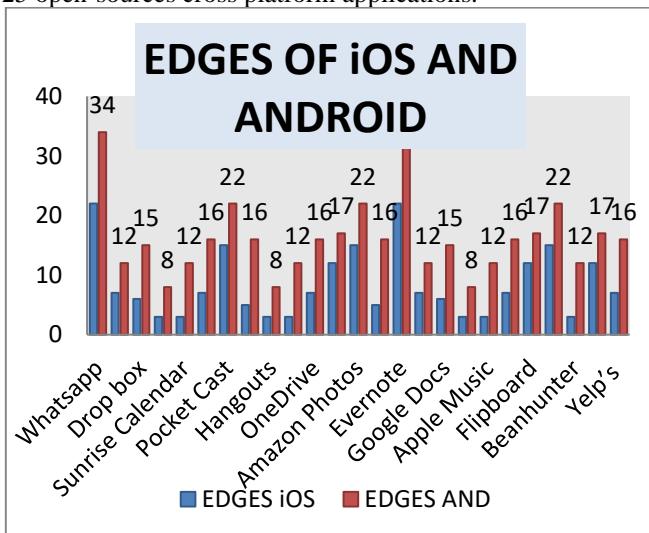


Fig. 3. Graph Representations of Edges

Fig 3 shows the graphical representation of the edges extracted from the applications of iOS and Android.

Table- I: Traces and Actions of iOS and Android

S.No	Name of cross platform apps	Type	EDGES		UNIQUE STATES	
			iOS	AND	iOS	AND
1	Whatsapp	Chat	22	34	14	17
2	Facebook messenger	Messaging	7	12	4	11
3	Drop box	Upload files	6	15	5	13
4	Wunderlist	List and Notes	3	8	5	11
5	Sunrise Calendar	Calendar	3	12	7	8
7	Microsoft Word	Documents	7	16	5	14
8	Spotify	Music	12	17	9	10
9	Pocket Cast	Service apps	15	22	8	9
10	IF	Update Photos	5	16	5	12
11	Hangouts	Access files	3	8	5	11
12	Google Drive	changes to a file	3	12	7	8
13	OneDrive	Word or Excel	7	16	5	14
14	Google Photos	photo apps	12	17	9	10
15	Amazon Photos	Compression	15	22	8	9
16	Google Keep	reminders	5	16	5	12
17	Evernote	Upload documents	22	34	14	17
18	Google Calendar	Upcoming agenda	7	12	4	11
19	Google Docs	Microsoft's DOCX	6	15	5	13
20	Google Play Music	Online radio	3	8	5	11
21	Apple Music	Music catalog	3	12	7	8
22	News Republic	Latest headlines	7	16	5	14
23	Flipboard	News Republic	12	17	9	10
24	Four Square	Local reviews	15	22	8	9
25	Bean hunter	Find cafes	3	12	7	8

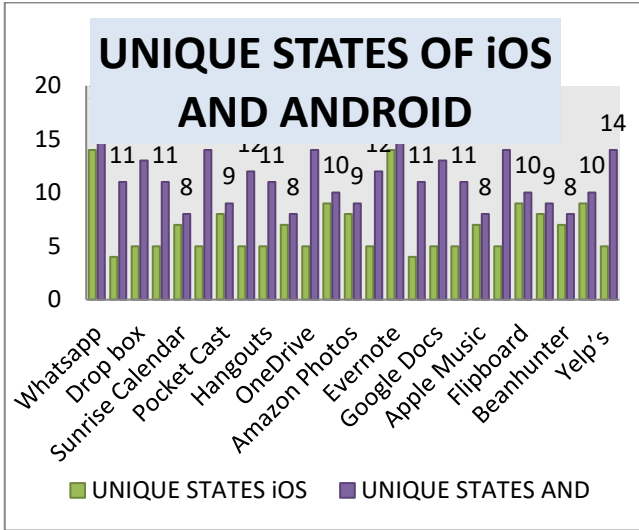


Fig. 4 Graph Representations of Unique States

The Fig 4 shows the Unique States evaluated from the given 25 open-source cross platform application. The graph clearly shows the difference in the states of both iOS and Android.

Table- II: Bug Severity Descriptions

Severity	Description	level
Critical	no loss in features, no work-around	5
Major	no loss in features, with possible work-around	4
Normal	Makes a task difficult to use data	3
Minor	Functionality not affecting, behavior is not expected	2
Trivial	Functionality not affecting, surface problem	1

Table II gives the descriptions about the bug severity. The majority of the false encouraging points in the detailed data inconsistencies are expected to the UI configuration state which has been implemented differently on both the platforms.

IV. CONCLUSION

The need for cross platform mobile application development has become an undeniably regular industry practice. In order to make the developed application consistent across multiple platforms is very important. So there is need for mobile developers and analyzers to test the application's consistency and guarantee that the behavior of the

application is same across multiple platforms. In this work, the implemented featCHECK technique consequently identifies and visualizes inconsistencies among iOS and Android versions which have same mobile applications

The performance valuation on 25 application couples demonstrates that the GUI based perfect could offer a good solution and also it maps the application-pairs.

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