

## Disclosure of Positioning Misrepresentation for Portable Application

**Nirupam Samanta**

M.Tech Student

Aurora's Scientific, Technological and Research Academy.

**Ms. S. Shalini**

Assistant Professor

Aurora's Scientific, Technological and Research Academy.

### **Abstract:**

*Positioning extortion in the versatile App advertises alludes to fake or beguiling exercises which have a reason for knocking up the Apps in the prominence list. In fact, it turns out to be increasingly visit for App designers to utilize shady means, for example, swelling their Apps' deals or posting fake App evaluations, to confer positioning misrepresentation. While the significance of anticipating positioning misrepresentation has been generally perceived, there is constrained comprehension and research around there. To this end, in this paper, we give a comprehensive perspective of positioning extortion and propose a positioning misrepresentation identification framework for versatile Apps. In particular, we initially propose to precisely find the positioning misrepresentation by mining the dynamic time frames, specifically driving sessions, of portable Apps. Such driving sessions can be utilized for identifying the neighborhood abnormality rather than worldwide irregularity of App rankings. Moreover, we explore three sorts of confirmations, i.e., positioning based confirmations, rating based confirmations and audit based confirmations, by demonstrating Apps' positioning, rating and survey practices through factual speculations tests. Also, we propose an improvement based total strategy to incorporate every one of the confirmations for misrepresentation recognition. At last, we assess the proposed framework with genuine App information gathered from the iOS App Store for quite a while period. In the trials, we approve the adequacy of the proposed framework, and demonstrate the versatility of the recognition calculation and in addition some normality of positioning extortion exercises.*

**Keywords:** Mobile Apps, Ranking Fraud Detection, Evidence Aggregation, Historical Ranking Records, Rating and Review, Recommendation app, KNN.

### **INTRODUCTION**

The quantity of mobile Apps has developed at an amazing rate in the course of recent years. For instances, the growth of apps were increased by 1.6 million at Apple's App store and Google Play. To increase the development of mobile Apps, many App stores launched daily App leader boards, which demonstrate the chart rankings of most popular Apps. Indeed, the App leader board is one of the most important ways for promoting mobile Apps. A higher rank on the leader board usually leads to a huge number of downloads and million dollars in revenue. Therefore, App developers tend to explore various ways such as advertising campaigns to promote their Apps in order to have their Apps ranked as high as possible in such App leader boards.

However, as a recent trend, instead of relying on traditional marketing solutions, shady App developers resort to some fraudulent means to deliberately boost their Apps and eventually manipulate the chart rankings on an App store. This is usually implemented by using so called "bot farms" or "human water armies" to inflate the App downloads, ratings and reviews in a very short time[10]. There are some related works, for example, web positioning spam recognition, online survey spam identification and portable App suggestion, but the issue of distinguishing positioning misrepresentation for mobile Apps is till under investigated.

The problem of detecting ranking fraud for mobile Apps is still underexplored. To overcome these essentials, in this paper, we build a system for positioning misrepresentation discovery framework for portable apps that is the model for detecting ranking fraud in mobile apps. For this, we have to identify several important challenges. First, fraud is happen any time during the whole life cycle of app, so the identification of the exact time of fraud is needed. Second, due to the huge number of mobile Apps, it is difficult to manually label ranking fraud for each App, so it is important to automatically detect fraud without using any basic information. Mobile Apps are not always ranked high in the leader board, but only in some leading events ranking that is fraud usually happens in leading sessions. Therefore, main target is to detect ranking fraud of mobile Apps within leading sessions. First propose an effective algorithm to identify the leading sessions of each App based on its historical ranking records.

Then, with the analysis of Apps' ranking behaviors, find out the fraudulent Apps often have different ranking patterns in each leading session compared with normal Apps. Thus, some fraud evidences are characterized from Apps' historical ranking records. Then three functions are developed to extract such ranking based fraud evidences. Therefore, further two types of fraud evidences are proposed based on Apps' rating and review history, which reflect some anomaly patterns from Apps' historical rating and review records. In addition, to integrate these three types of evidences, an unsupervised evidence-aggregation method is developed which is used for evaluating the credibility of leading sessions from mobile Apps.

#### **EXISTING SYSTEM:**

- In the literature, while there are some related work, such as web ranking spam detection, online review spam detection and mobile App recommendation, the problem of detecting ranking fraud for mobile Apps is still under-explored.

- Generally speaking, the related works of this study can be grouped into three categories.
- The first category is about web ranking spam detection.
- The second category is focused on detecting online review spam.
- Finally, the third category includes the studies on mobile App recommendation

#### **DISADVANTAGES OF EXISTING SYSTEM:**

- Although some of the existing approaches can be used for anomaly detection from historical rating and review records, they are not able to extract fraud evidences for a given time period (i.e., leading session).
- Cannot able to detect ranking fraud happened in Apps' historical leading sessions
- There is no existing benchmark to decide which leading sessions or Apps really contain ranking fraud.

#### **PROPOSED SYSTEM:**

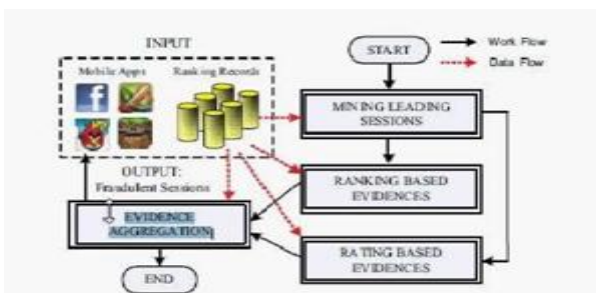
- We first propose a simple yet effective algorithm to identify the leading sessions of each App based on its historical ranking records. Then, with the analysis of Apps' ranking behaviors, we find that the fraudulent Apps often have different ranking patterns in each leading session compared with normal Apps. Thus, we characterize some fraud evidences from Apps' historical ranking records, and develop three functions to extract such ranking based fraud evidences.
- We further propose two types of fraud evidences based on Apps' rating and review history, which reflect some anomaly patterns from Apps' historical rating and review records.
- In Ranking Based Evidences, by analyzing the Apps' historical ranking records, we observe that Apps' ranking behaviors in a leading event always satisfy a specific ranking pattern, which consists of three different ranking phases, namely, rising phase, maintaining phase and recession phase.

- In Rating Based Evidences, specifically, after an App has been published, it can be rated by any user who downloaded it. Indeed, user rating is one of the most important features of App advertisement. An App which has higher rating may attract more users to download and can also be ranked higher in the leaderboard. Thus, rating manipulation is also an important perspective of ranking fraud.
- In Review Based Evidences, besides ratings, most of the App stores also allow users to write some textual comments as App reviews. Such reviews can reflect the personal perceptions and usage experiences of existing users for particular mobile Apps. Indeed, review manipulation is one of the most important perspectives of App ranking fraud.

**ADVANTAGES OF PROPOSED SYSTEM:**

- The proposed framework is scalable and can be extended with other domain generated evidences for ranking fraud detection.
- Experimental results show the effectiveness of the proposed system, the scalability of the detection algorithm as well as some regularity of ranking fraud activities.
- To the best of our knowledge, there is no existing benchmark to decide which leading sessions or Apps really contain ranking fraud. Thus, we develop four intuitive baselines and invite five human evaluators to validate the effectiveness of our approach Evidence Aggregation based Ranking Fraud Detection (EA-RFD).

**SYSTEM ARCHITECTURE**



**Fig 1. The frame work of the Ranking fraud detection system for Mobile Apps**

With the increase in the number of web Apps, to detect the fraud Apps, this paper proposes a simple and effective system. Fig.1 shows the Framework of Fraud ranking discovery in mobile app

**Module 1: Leading events**

Given a positioning limit  $K \in [1, K]$  a main occasion e of App a contains a period range also, relating rankings of a, Note that positioning edge  $K^*$  is applied which is normally littler than  $K$  here on the grounds that  $K$  may be huge (e.g., more than 1,000), and the positioning records past  $K$  (e.g., 300) are not exceptionally helpful for recognizing the positioning controls. Moreover, it is finding that a few Apps have a few nearby driving even which are near one another and structure a main session.

**Module 2: Leading Sessions**

Instinctively, mainly the leading sessions of mobile app signify the period of popularity, and so these leading sessions will comprise of ranking manipulation only. Hence, the issue of identifying ranking fraud is to identify deceptive leading sessions. Along with the main task is to extract the leading sessions of a mobile App from its historical ranking records.

**Module 3:**

Identifying the leading sessions for mobile apps Basically, mining leading sessions has two types of steps concerning with mobile fraud apps. Firstly, from the Apps historical ranking records, discovery of leading events is done and then secondly merging of adjacent leading events is done which appeared for constructing leading sessions. Certainly, some specific algorithm is demonstrated from the pseudo code of mining sessions of given mobile App and that algorithm is able to identify the certain leading events and sessions by scanning historical records one by one.

**Module 4:**

Identifying evidences for ranking fraud detection

### **Ranking Based Evidence**

It concludes that leading session comprises of various leading events. Hence by analysis of basic behaviour of leading events for finding fraud evidences and also for the app historical ranking records, it is been observed that a specific ranking pattern is always satisfied by app ranking behavior in a leading event.

### **Rating Based Evidence**

Previous ranking based evidences are useful for detection purpose but it is not sufficient. Resolving the problem of “restrict time reduction”, identification of fraud evidences is planned due to app historical rating records. As we know that rating is been done after downloading it by the user, and if the rating is high in leaderboard considerably that is attracted by most of the mobile app users. Spontaneously, the ratings during the leading session gives rise to the anomaly pattern which happens during rating fraud. These historical records can be used for developing rating based evidences.

### **Review Based Evidence**

We are familiar with the review which contains some textual comments as reviews by app user and before downloading or using the app user mostly prefer to refer the reviews given by most of the users. Therefore, although due to some previous works on review spam detection, there still issue on locating the local anomaly of reviews in leading sessions. So based on apps review behaviors, fraud evidences are used to detect the ranking fraud in Mobile app.

### **CONCLUSION**

This paper, gives the ranking fraud detection model for mobile apps. Now days many of mobile app developers uses various frauds techniques to increase their rank. To avoid this, there are various fraud detection techniques which are studied in this paper. We detect the ranking fraud using actual fraud reviews. This paper proposes the time efficient system to detect the fraud Apps.

### **REFERENCES**

- [1] B. Zhou, J. Pei, and Z. Tang. A spamicity approach to web spam detection. In Proceedings of the 2008 SIAM International Conference on Data Mining, SDM'08, pages 277–288, 2008.
- [2] A. Ntoulas, M. Najork, M. Manasse, and D. Fetterly. Detecting spam web pages through content analysis. In Proceedings of the 15th international conference on World Wide Web, WWW '06, pages 83–92, 2006.
- [3] N. Spirin and J. Han. Survey on web spam detection: principles and algorithms. SIGKDD Explor. Newsl., 13(2):50–64, May 2012.
- [4] E.-P. Lim, V.-A. Nguyen, N. Jindal, B. Liu, and H. W. Lauw. Detecting product review spammers using rating behaviors. In Proceedings of the 19th ACM international conference on Information and knowledge management, CIKM '10, pages 939–948, 2010.
- [5] Z.Wu, J.Wu, J. Cao, and D. Tao. Hysad: a semisupervised hybrid shilling attack detector for trustworthy product recommendation. In Proceedings of the 18th ACM SIGKDD international conference on Knowledge discovery and data mining, KDD '12, pages 985–993, 2012
- [6] Getjar mobile application recommendations with very sparse datasets. K. Shi and K. Ali. In Proceedings of the 18th ACM SIGKDD international conference on Knowledge discovery and data mining, KDD '12, pages 204–212, 2012.
- [7] Ranking fraud Mining personal context- aware preferences for mobile users. H. Zhu, E. Chen, K. Yu, H. Cao, H. Xiong, and J. Tian. In Data Mining (ICDM), 2012 IEEE 12th International Conference on, pages 1212–1217, 2012.



[8] detection for mobile apps H. Zhu, H. Xiong, Y. Ge, and E. Chen. A holistic view. In Proceedings of the 22nd ACM international conference on Information and knowledge management, CIKM '13, 2013.

[9] Exploiting enriched contextual information for mobile app classification, H. Zhu, H. Cao, E. Chen, H. Xiong, and J. Tian. In Proceedings of the 21st ACM international conference on Information and knowledge management, CIKM '12, pages 1617–1621, 2012.

[10] spammers using behavioral Footprints A. Mukherjee, A. Kumar, B. Liu, J. Wang, M. Hsu, M. Castellanos, and R. Ghosh. In Proceedings of the 19th ACM SIGKDD international conference on Knowledge discovery and data mining, KDD '13, 2013.

[11] Detecting product review spammers using rating behaviors. E.-P. Lim, V.-A. Nguyen, N. Jindal, B. Liu, and H. W. Lauw In Proceedings of the 19th ACM international conference on Information and knowledge management, CIKM '10, pages 939–948, 2010.