

Method to Study and Analyze Fraud Ranking In Mobile Apps



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Abstract:

Ranking fraud in the mobile App market refers to false or deceptive activities which have a reason of bumping up the Apps in the popularity list. Certainly, it becomes more and more frequent for App developers to use shady means, such as inflating their Apps' sales or posting phony App ratings, to commit ranking fraud. While the importance of preventing ranking fraud has been widely recognized, there is limited understanding and research in this area.

A ranking fraud detection system for mobile Apps was developed. Specifically, this ranking fraud happened in leading sessions and provided a method for mining leading sessions for each App from its historical ranking records and identified ranking based evidences, rating based evidences and review based evidences for detecting ranking fraud. Moreover, we proposed an optimization based aggregation method to integrate all the evidences for evaluating the credibility of leading sessions from mobile Apps. A unique perspective of this approach is that all the evidences can be modeled by statistical hypothesis tests, in this paper we want to propose more effective fraud evidences and analyze the latent relationship among rating, review and rankings. Moreover, we will extend our ranking fraud detection approach with other mobile App related services, such as mobile Apps recommendation, for enhancing user experience.

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 still 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 underexplored.

- 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

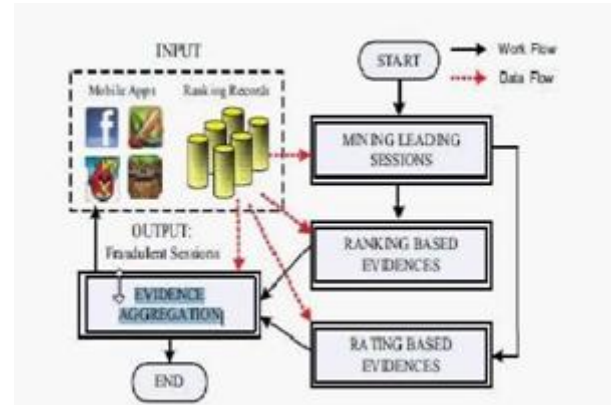


Fig 1. The frame work of the Ranking fraud detection system for Mobile Apps

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 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.

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