Mobile App Recommendation Based on Rating Review & Ranking
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Abstract:
Now a day’s everyone is using smart phones. There is need of various applications to be installed on smart phone. To download application smart phone user has to visit play store such as Google Play Store, Apples store etc. When user visit play store then he is able to see the various application lists. This list is built on the basis of promotion or advertisement. User doesn’t have knowledge about the application (i.e. which applications are useful or useless). So user looks at the list and downloads the applications. But sometimes it happens that the downloaded application won’t work or not useful. That means it is fraud in mobile application list. To avoid this fraud, we are making application in which we are going to list the applications. To list the application first we are going to find the active period of the application named as leading session. We are also investing the three types of evidences: Ranking based evidence, Rating based evidence and Review based evidence. Using these three evidences finally we are calculating aggregation. We evaluate our application with real world data collected form play store for long time period. In the experiments, we validate the effectiveness of the proposed system, and show the scalability of the detection algorithm as well as some regularity of ranking fraud activities.

Keywords: Mobile Apps, ranking fraud detection, historical ranking records, evidence aggregation, review, ranking and rating.

I. INTRODUCTION
As smartphones emerges new technologies like android and iOS operating system took a boost in market. Mobile application started growing at such a high rate. As a study says millions of apps are there on apple’s app store and on Google Play. This started a new business in computer world and became a reason to earn thousands of dollars and downloads. Daily leader board is published by these markets contains the most popular apps which will consequently be downloaded and rated most high by users. Some developers may use some marketing strategies like an advertisement campaign for promotion of their app. However this part of technology is also not safe from threats. Mobile app market, we refer it as market, is manipulated by some fraudulent app developers to bump up their app high in the rank list, as an app in leader board confirms high downloads and high income. Shady means are used to make such a fraud and implemented using “bot farms” which is also called “Human water armies”. In this area some literature survey is there, for example, spam detection for web ranking, mobile app recommendations, and some online review based spam detection. Our study thus focuses on an integrated approach, for various evidences, to find Mobile App ranking fraud and also recommend the most relevant App that is most genuine. For this we have to go through challenges like first we need to find at what time the fraud is happening it means exact time of fraud is needed. Secondly we know that there is tremendous number of Apps present in market so it is nearly impossible to physically mark ranking fraud for every App, so it’s crucial to automatically distinguish fraud without utilizing any essential data. Mobile Apps are not commonly ranked high in the leader board, but instead just in a few events ranking frauds more often than not happens in leading sessions. In this way, fundamental target is to recognize ranking fraud of mobile Apps inside of leading sessions. Initially propose an efficient algorithm to recognize the main sessions of every App depends on its previous ranking records. By then, with the examination of Apps’ ranking practices, find the fake Apps consistently have unique ranking examples in every leading session contrasted with ordinary Apps. Along these lines, some fraud confirmations are portrayed from Apps’ previous ranking records. By then three limits are produced to concentrate such ranking based fraud confirmations. Thusly, help two kinds of fraud confirmations are proposed taking into account Apps’ rating and survey history, which mirror some inconsistency patterns from Apps’ previous rating and review records. Also, to coordinate these three kinds of unsupervised proof collection procedure is created which is used for assessing the validity of leading sessions from mobile Apps.

II. LITERATURE SURVEY
In this section discuss existing work done by the researchers for text mining process. In paper [1], author made ranking fraud location framework for mobile Apps. Specifically, they at first showed that ranking fraud happened in leading sessions and gave a strategy to mining leading sessions for every App from its previous ranking records. By then, they recognized ranking based confirmations, rating based proofs and review based proofs for distinguishing ranking fraud. They likewise proposed a optimization based collection strategy to consolidate each one of the proofs for surveying the validity of leading sessions from mobile Apps. In paper [2], author have focused on various parts of substance build spam regarding the Web and showed different heuristic schedules for recognizing content based spam. Here, they continue with examinations of “web spam”: the infusion of misleadingly made pages into the web with a particular deciding objective to affect the results from web crawlers, to direct individuals to particular pages for the purpose of excitement or advantage. In paper [3], author has reported a review on Web spam location, which altogether exhibits the rules and algorithm in the literature. Undoubtedly, the work of Web ranking spam identification is principally in
light of the examination of ranking measures of web searchers, for instance, Page Rank and question term frequency. This is not the same as ranking fraud location for mobile Apps. They sort each present algorithm into three classifications in light of the type of information they use: content-based techniques, link based strategies and techniques on the basis of non-conventional information, for instance, customer behavior, snaps, and HTTP sessions. In paper [4], authors have seen a couple of representative behaviors of review spammers and model these practices to identify the spammers. This paper expects to perceive clients creating spam overviews or review spammers. They perceive a couple trademark practices of review spammers and model these practices with a specific end goal to recognize the spammers. Authors attempt to exhibit the going with practices. In any case, spammers may target specific things or item stores up in order to grow their impact. Second, they have a tendency to go out of order from exchange experts in their assessments of items. In paper [5], authors have analyzed the issue of discovering hybrid shilling attacks on rating information. The methodology depends on can be utilized for reliable item suggestion and the semi-supervised learning. This paper displays a Hybrid Shilling Attack Detector or HySAD for short, to handle this issue. Specifically, HySAD familiarizes MC-Relief with select successful recognition metrics and Semi managed Naïve Bays (SNB) to correctly isolate Random-Filler model aggressors and Average- Filler model attackers from standard clients. In paper [6], authors have analyzed the issue of singleton survey spam detection. Specifically, they handled this issue by recognizing the co-anomaly pattern in different review based time arrangement. Also some of above strategies can be used for anomaly detection from previous rating and overview records, they are not prepared to focus fraud evidences for a given time period (i.e., leading session). In paper [7], author created a mobile App recommender framework, Appjoy, which depends on user’s App use records to assemble an inclination matrix despite utilizing explicit client ratings. In paper [8], author analyzed a few suggestion models and proposed a content-based collaborative separating model, called Eigenapp, for prescribing Apps in their Web website Getjar. Also, a few researchers analyzed the issue of misusing advanced logical data for mobile App suggestion.

III. PROBLEM OF ANALYSIS

The Mobile App is a very popular and well known concept due to the rapid advancement in the mobile technology. Due to the large number of mobile Apps, ranking fraud is the key challenge in front of the mobile App market. There are millions of apps are available in market for the application of mobile users. User doesn’t have knowledge about the application (i.e. which applications are useful or useless). So user looks at the list and downloads the applications. But sometimes it happens that the downloaded application won’t work or not useful. That means it is fraud in mobile application list. To avoid this fraud, we are making application in which we are going to list the applications. In this paper, we propose a brief view of ranking fraud and propose a ranking fraud detection system for mobile Apps. Specifically, we first propose to accurately locate the ranking fraud by mining the active periods by using mining leading session algorithm. Furthermore, we investigate three types of evidences, i.e., ranking based evidences, rating based evidences and review based evidences, by studying historical records. We used an optimal aggregation method to integrate all the evidences for fraud detection. Finally, we evaluate the proposed system with real-world App data collected from the Google App Store for a long time period. In the experiments, we validate the effectiveness of the proposed system, and show the scalability of the detection algorithm as well as some regularity of ranking fraud activities. To list the application first we are going to find the active period of the application named as leading session. We are also investing the three types of evidences are as follows:

**Ranking Based Evidences**: 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.

**Rating Based Evidences**: 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 leader board. Thus, rating manipulation is also an important perspective of ranking fraud.

**Review Based Evidences**: 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.

IV. PROPOSED WORK

As there is increasing in the number of mobile apps. Fraud ant apps must be detected. To avoid the fraud we are going to list the applications. We provide a brief view of ranking fraud and propose a ranking fraud detection system for mobile apps. We have proposed a simple and effective algorithm for detecting fraud ranking and generating ranking and review.

4.1 For detecting ranking fraud

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R = Review set
P = Positive/Negative words dictionary
fb = Fraud detection bit
rt = Rating of app
rk = Ranking of app
for each i ∈ (1, R) ; i++
do
Split i ;
Add i to S[i]; j++
do
Search j in P;
Set count; return count
if (count >0)
Set sentiment = 1;
else if (count <0)
set sentiment = -1;
else if (count = 0)
set sentiment =0;
return sentiment
if (sentiment>0)
if (rt>=4)
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then set fb = 0;
else set fb = 1;
else if (sentiment < 0)
if (rt <= 2)
then set fb = 0;
then set fb = 1;
else if (sentiment != 0)
if (rt <= 4 And rt >= 2)
then set fb = 0;
else set fb = 1;
End

4.2 Generating rating and review

rk(new) = New Ranking of App
rt(old) = Old Ranking of App
rt(new) = New Rating of App
A = Set of All Apps

for each i ∈ (1, A); i++
do
rt(new)++ = avg (rt (old) [i]);
[Where fb=0]
returnrt (new);
sortrt(new)in Decending order
for each j ∈ (1,rt(new));j++
do
rk(new)=j;
returnrk(new);

Proposed work includes the following –

- Behavior of Review Spammer:- We will detect users generating spam reviews or review spammers. We identify several characteristic behaviors of review spammers and model these behaviors so as to detect the spammers. In particular, we seek to model the following behaviors. First, spammers may target specific products or product groups in order to maximize their impact. Second, they tend to deviate from the other reviewers in their ratings of products. I propose scoring methods to measure the degree of spam for each reviewer and apply them on review dataset. We finally show that the detected spammers have more significant impact on ratings compared with the unhelpful reviewers.

- Ranking Aggregation: -We are going to detect the ranking fraud mobile app in rank aggregation on the basis of review and rating evidence by using historical dataset and my own dataset.

- The Experimental Data: -The experimental data sets were collected from the Apple’s App Store (U.S.) from February 2, 2010 to September 17, 2012. The data sets contain the daily chart rankings1 of Apps. Furthermore, each data set also contains the user ratings and review information.

- The proposed system contributes the new concept of recommendation system for the mobile applications to the number of users. The recommendation system works on the number of reviews and ratings are given by the users for the specific product. The system implement the recommendation system that restricts the fake reviews. The system recommendation has been generated for the better results to the user on the basis of previous records. Also eliminates the fake reviews from the dataset and detect the web rank. The system saves the time as well as memory.

- The system shows graph of genuine application rating.

This is stage of the project when the theoretical design is turned out in working system. Thus it can be considered to be the most critical stage in achieving a successful new system and in giving the user, confidence that the new system will work and be effective. The framework involves careful planning, investigation of existing system and its constraints on implementation, designing of methods to achieve goal. In the first step system will extract historical data from the dataset. It will compare pair wise matching the ratings and reviews. In the second stage user will singing into the system with the help of specified username and password. Then user will provide the ratings and reviews to the specified data. If there will be a fraud detected then it will predict rating and reviews, keeping old rating and reviews. After that it will merge both of the data. In third stage it will provide output. It recommends users that apps are in the top 3 ratings or not. The system shows graph of genuine application ratings.

Figure 1. System architecture

V. CONCLUSION

In this way, we give the ranking fraud detection model for mobile apps. Now a day’s many of mobile app developers uses various frauds techniques to increase their rank. To avoid this, there are various fraud detection techniques. We developed a ranking fraud detection system for mobile Apps. Specifically, we first showed that ranking fraud happened in leading sessions and provided a method for mining leading sessions for each App from its historical ranking records. Then, we identified ranking based evidences, rating based evidences and review based evidences for detecting ranking fraud. Moreover, we proposed a mining Leading session algorithm for obtain mining leading session and aggregation method. In the future, we plan to study 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. We detect the ranking fraud using actual fraud reviews. Finally the system shows graph of genuine application rating.
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VII. REFERENCES

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