

A Novel Recommendation System for Mobile Apps Based on Detection of Ranking Frauds

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ABSTRACT

Recently, Smartphone gains more and more popularity not because of its look and feel but peoples are addicted to use various applications of Smartphone. With the use of smartphone apps user also worried to reveal their identity while downloading and installing app in their mobile as many apps may require user information. For example, what's app messenger, hika messenger etc uses contact list information from user cell phone, Google map, drive apps like uber, Ola etc uses contacts, location information of user. Mobile apps mostly varied and less understood typically for their popular functionality. Users are neglecting to install apps as security and privacy point of view. To overcome security issues there is need of such system which gives recommended app list to the end user. User recommendation is based on popularity as well as security concern.

Keyword: - Mobile Apps, Recommender Systems, Security and Privacy.

1. INTRODUCTION

Generally, mobile apps are developed using computer programming languages such as, java, android, .net, X-code etc. Apps are designed in such way to match with device layout such as computer, mobile, tabs etc. After successful completion of application programs is using programming language, they have to plot into electronic devices like, mobile, tabs, android wear etc. Application distributed platform is used to relate device and develop apps. In digital market, there are lots of mobile apps are available, few of them are freeware and at the same time others are In-purchased apps. Usually, apps are downloaded and installed in target device such as, android, windows, iPhone etc. The app availability based on the public demand, so the designer tools herd fast and rapid extension of mobile app into various other kinds such as games in mobile, factory computerization, services based on

GPS and based on location, banking, ticket purchasing, etc. There is a challenging issue in mobile app recommendation, because of sudden increase in quantity and the variety of mobile apps which in turn led to the conception of broad range of review and creation sources including blogs, magazines and online app services [1]. Each and every day multiple app stores launched a daily app leader boards that outputs the app ranking chart as per app popularity. Leader board is one of the most popular ways to promote mobile apps [3]. Therefore, according to the ranking on leader board app get downloaded by mobile users. And this is the reason of more advertising of mobile apps by their developers. However, leafy App developers find some fraudulent means to purposely boost their Apps which eventually manipulate the chart rankings on an App store. Mobile app ranking is varied according to its functionality with respect to privacy and security [2].

To provide better experience for end user some apps used personal information of user such as, contact list, location based service to extract user's current location. For instance, users may not expect to share their location as well as their other personal information. According to recent survey, it seems that multiple peoples were not interested to download the apps that required personal information of user and also many of them were uninstalling apps from their mobile phones which have functionality to access private information of user [4-6]. Therefore, there is a need of such a kind of system that provides recommendation of apps for end user with security and privacy perspective. We are going to propose a

recommendation system which work against user privacy and secrecy and then recommend apps for end user. For implementation such kind of system we will have to evaluate risk score i.e. privacy factor. It may be security wise and popularity wise score [2].

However, for security and popularity of apps we need to explore app dataset and required to refine it. And have to evaluate certain kind of permissions as, normal permissions having minimal risk to other applications. Dangerous permissions provides access for private data and finally signature/System permissions which required system signature certifications like, ability to control the overall process of the system. In app mining process we give preferences for rating based apps, review based app and also trying for survey based customer reviews [7]. Which can further combine together for generate recommended app-list for end user. By evaluation of risk score and app mining method we will have to prove efficiency and capability of our system. Our final contribution is detection of ranking fraud detection and app recommendation.

2. LITERATURE SURVEY

In the year 2008 B Zhou, J.Pie and Z.Tang published paper titled "A spamicity approach to web spam detection". In that paper he differentiated the spam mail detection as how we classify the mail as spam or ham. By using the stop words or special words which generally occur in the mail so we can sort them by those words [1]

In year 2009 W. Enck, M. Ontang, and P. McDaniel published paper titled On lightweight mobile phone application certification. In that paper the proceedings of the 16th ACM conference on Computer and communications security was focused on the security level which should be set for apps.[2]

In the year 2011 A.P. Felt, E. Chin, S. Hanna, D. Song, and D.Wagner published Android permissions demystified. In that paper proceedings of the 18th ACM conference on Computer and communications Security which was purely based on the security issues. [3]

In the year 2012 H. Zhu, H. Cao, E. Chen, H. Xiong, and J. Tian published paper on Exploiting enriched contextual information for mobile app classification. In that paper he made the classification of mobile app depending on there work.[4]

In the year 2013 S. Xie, G. Wang, S. Lin, and P. S. Yu published paper on Review spam detection via temporal pattern discovery which broadly was based on the review detection.[5]

In the year 2013 Matthijs.J Warrens published a paper on Comparison of Cohens Kappa in which he made the aggregation of fraud detection possible[6].

In the year 2015 Hengshu Zhu, Hui Xiong published paper on Discovery of Ranking Fraud for Mobile Apps in which he classified the apps as ranking, rating and review based and found out the leading sessions.[7]

3. PROPOSED SYSTEM

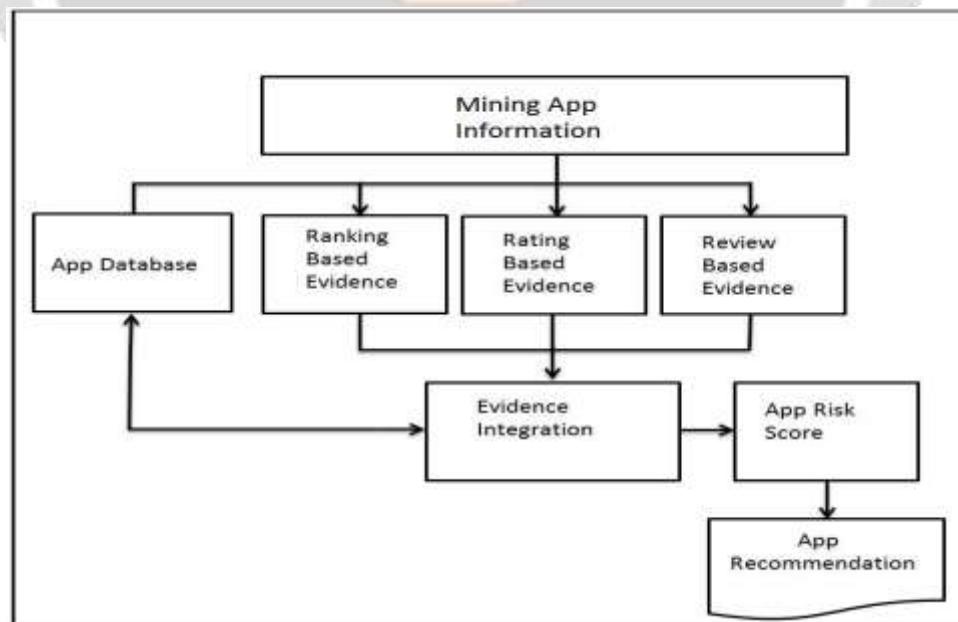


Fig -1. Architecture

3.1. App Mining:

App mining is the procedure of extracting technical, business and customer review and rating information for analysis and prediction i.e. to capture uncertain events. In system architecture diagram app mining method is divided into three parts such as, ranking based, rating based, survey based. It is described as follow:

Ranked based: It contains app information having certain rise and falls in their popularity.

Rating Based: It recognizes fake ratings about app i.e. random comments of users. It contains over wise popularity apps. For more download developer provides some discount or over some gift for users where user are attracted and gives positive comments regarding app.

Review based: Comment wise evidence. In this fraud detection is takes place using stop words to remove it. Basically normalized vector is used to calculate similarity between app functionalities.

3.2. Evidence Integration:

In this the result which is generated through ranking, rating and survey is considered and then it is integrated and that information is passed for calculating the risk score.

3.3. Calculation of Risk Score:

Two types of principle are involved to calculate risk score:

Security principle: App risk score is evaluated and arranged it in ascending order. If risk score is similar of more than one app then it is evaluated or ranked by popularity score.

Popularity principle: Ranking apps are evaluated as per popularity score and arrange it in descending order. If risk score is similar of more than one app then it is evaluated or ranked by security score.

3.4. App Recommendation:

After all processing done regarding security and popularity issues of mobile apps, final stage is to recommend sorted apps to the end user. So, that user can satisfy their need and worries About app usage.

In our system we use four major components: App Category List, Recommendation List, Search Application, and Update Profile.

1] App Category List: The app category consists of the apps which are fetched from the Google play store and the apps which the user contains in his mobile.

2] Recommendation Application: The recommendation application consists of two types of notification list: First according to the user's age, profession and gender. Second consist of notifying to uninstall the apps which are no longer used by the user.

3] Search Application: It consists of the link of Google play store so that user can download the app according to his/her interest.

4] Update Profile: It contains the user data like age, profession and gender which is used to recommend the app.

4. ALGORITHM

4.1 Ranking Algorithm:

- Input** 1: a's historical ranking records R_a ;
 2: the ranking threshold K^* ;
 3: the merging threshold \emptyset ;

Output: the set of a's leading sessions S_a ;

Initialization:

- $S_a = \emptyset$; 1: $E_s = \emptyset$; $e = \emptyset$; $s = \emptyset$; $t_{start}^e = 0$;
 2: for each $i \in [1, |R_a|]$ do
 3: if $r_i^a \leq K^*$ and $t_{start}^e = 0$ then
 4: $t_{start}^e = t_i$;
 5: else if $r_i^a > K^*$ and $t_{start}^e \neq 0$ then
 6: $t_{end}^e = t_{i-1}$; $e = \langle t_{start}^e, t_{end}^e \rangle$;
 7: if $E_s = \emptyset$ then
 8: $E_s \cup = e$; $t_{start}^s = t_{start}^e$; $t_{end}^s = t_{end}^e$;
 9: else if $(t_{start}^e - t_{end}^s) < \emptyset$ then
 10: $E_s \cup = e$; $t_{end}^s = t_{end}^e$;
 11: else then
 12: $s = \langle t_{start}^s, t_{end}^s, E_s \rangle$;
 13: $S_a \cup = s$; $s = \emptyset$ is a new session;
 14: $E_s = \{e\}$; $t_{start}^s = t_{start}^e$; $t_{end}^s = t_{end}^e$;
 15: $t_{start}^e = 0$; $e = \emptyset$ is a new leading event;
 16: return S_a ;



4.2 Fraud Detection Algorithm for Fake Reviews

Step 1: Comparing customer 1 reviews and customer 2 review.

Step 2: Then checking which one is more similar type of reviews exact review then from given reviews one review from that similar review is considered for a particular product.

Step 3: Keeping a criteria that only registered users can give review.

Step 4: Giving preference to top 10 reviews for a particular product that will help the users to decide whether to download that app or not.

5. MATHEMATICAL MODEL

Sys = { $\psi_1, \psi_2, \psi_3, \psi_4, \psi_5$ }

ψ_1 = Data collector app.

This data is collected from sample population decided at start of designing of system.

ψ_2 = Actual recommender application for end user that analyse user profile and recommend based on clusters

ψ_3 = Web based handshaking interfaces or services. Handshake between Android and web based app information is necessary

ψ_4 = Application for super user of the system.

ψ_5 = Application permission and security.

$\psi_1 = \{\psi_{1Ip}, \psi_{1Op}, \psi_{1Fn}\}$

Here, ψ_{1Ip} is Input,

ψ_{1Op} is Output and

ψ_{1Fn} is Function

WHERE,

$\psi_{1Ip} = \{\psi_{1Ip1}, \psi_{1Ip2}\}$

ψ_{1Ip1} = Add information of user belongs to sample population

ψ_{1Ip2} = Access Logs Details of selected sample user

$\psi_{1Fn} = \{\psi_{1Fn1}, \psi_{1Fn2}, \psi_{1Fn3}, \psi_{1Fn4}, \psi_{1Fn5}\}$

ψ_{1Fn1} = Save personnel information

ψ_{1Fn2} = Get installed app details

ψ_{1Fn3} = Filter system required app

ψ_{1Fn4} = Get access logs details as per application

ψ_{1Fn5} = Add these logs details after analyzing them

$\psi_{1Op} = \{\psi_{1Op1}\}$

ψ_{1Op1} = Get analytical results of added logs

2. $\Psi_4 = \{\psi_{4Ip}, \psi_{4Op}, \psi_{4Fn}\}$

$\psi_{4Ip} = \{\psi_{4Ip1}\}$

ψ_{4Ip1} = Manage / Generate .apk file in case of any changes in respective apps

$\psi_{4Fn} = \{\psi_{4Fn1}, \psi_{4Fn2}, \psi_{4Fn3}, \psi_{4Fn4}, \psi_{4Fn5}, \psi_{4Fn6}\}$

ψ_{4Fn1} = Get App Permissions using API

ψ_{4Fn2} = Add app permission

ψ_{4Fn3} = Update app permission list

ψ_{4Fn4} = Manage app details

ψ_{4Fn5} = Manage app category list

ψ_{4Fn6} = Generate Clusters based on logs information and permission matches and reviews.

$\psi_{4Op} = \{\psi_{4Op1}, \psi_{4Op2}, \psi_{4Op3}\}$

ψ_{4Op1} = Update App Permissions

ψ_{4Op2} = Update app permission list

ψ_{4Op3} = Update app category list

3. $\Psi_3 = \{\psi_{3Ip}, \psi_{3Op}, \psi_{3Fn}\}$

$\psi_{3Ip} = \{\psi_{3Ip1}, \psi_{3Ip2}, \psi_{3Ip3}, \psi_{3Ip4}\}$

ψ_{3Ip1} = Application set

ψ_{3Ip2} = Permission Set

ψ_{3Ip3} = Categories set

ψ_{3Ip4} = User request

$\psi_{3Fn} = \{\psi_{3Fn1}, \psi_{3Fn2}, \psi_{3Fn3}, \psi_{3Fn4}, \psi_{3Fn5}\}$

ψ_{3Fn1} = Extract web based feature

ψ_{3Fn2} = Web based features preprocessing

ψ_{3Fn3} = Get application category

ψ_{3Fn4} = User request web service

ψ_{3Fn5} = App recommendation list

$\psi_{3Op} = \{\psi_{3Op1}, \psi_{3Op2}\}$

ψ_{3Op1} = Classification of app

ψ_{3Op2} = Recommended App

4. $\Psi_2 = \{\psi_{2Ip}, \psi_{2Op}, \psi_{2Fn}\}$

$\psi_{2Ip} = \{\psi_{2Ip1}, \psi_{2Ip2}, \psi_{2Ip3}, \psi_{2Ip4}\}$

ψ_{2Ip1} = Personnel Details

ψ_{2Ip2} = Trendy Apps requests

ψ_{2Ip3} = Current Apps requests

ψ_{2Ip4} = Personnel App recommendation based on profile

$\psi_{2Fn} = \{\psi_{2Fn1}, \psi_{2Fn2}, \psi_{2Fn3}\}$

ψ_{2Fn1} = Get app recommendation request

$\psi 2Fn2$ = Get user profile based app details

$\psi 2Fn3$ = Get Trendy app list by calculating and analyzing the request

$\psi 2Op = \{\psi 2Op1, \psi 2Op2, \psi 2Op3\}$

$\psi 2Op1$ = List of current recommended application based on profile with category, review and risk score

$\psi 2Op2$ = List of trendy app

$\psi 2Op3$ = List of personalized recommendation app

6. EXPERIMENTAL RESULTS

We have taken top 20 paid apps and top 20 free apps for our analysis. We have maintained two databases first is the dynamic database where we have provided the link of the Google play store and we are continuously updating the database with the top trending free and paid apps. Second database is of the registered user, when the user registers himself and login to the app then all the apps data related to which apps user is using is stored in the database.

Also we made some synthetic data in which we are manually creating the fake ranks, reviews and ratings for some apps to check whether our system is detecting it as fraud app or not and recommending to uninstall. The app is providing recommendation in 3 types:

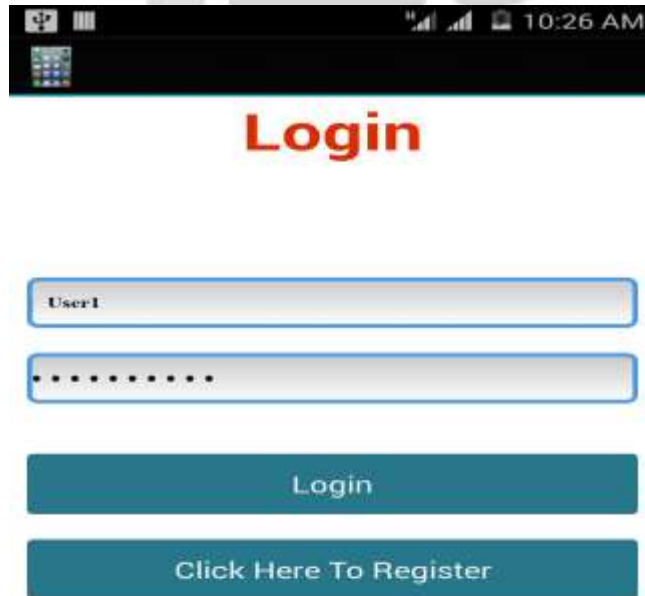
1. Using age, profession and gender the app is recommending those apps which are of user interest.
2. If a app is installed on user mobile and it is fraud or not in top 20 free or paid apps we are using the integration of ranking ,rating and review for that app and if all the condition are true then our app will recommend user to uninstall that suspicious or fraud app.
3. By using the app usage count if an app in user mobile is not used for longer time than it should notify the user to uninstall that app.

Using is analysis we have assumed following values:

1. For paid apps if the rate count of an app suddenly increases by **5000** every time then that app is suspicious app.
2. for free apps if the rate count of an app suddenly increases by **1, 50,000** every time then that app is suspicious app.
3. If there is a jump of 0.2 in rating for an app then that app is considered as suspicious.

6.1 Screenshots:

6.1.1 Login



6.1.2 Registration Page

Connected as a media device

RegistrationActivity

poonam

falley

nsk

8149642690

abc@gmail..com

Male

Female

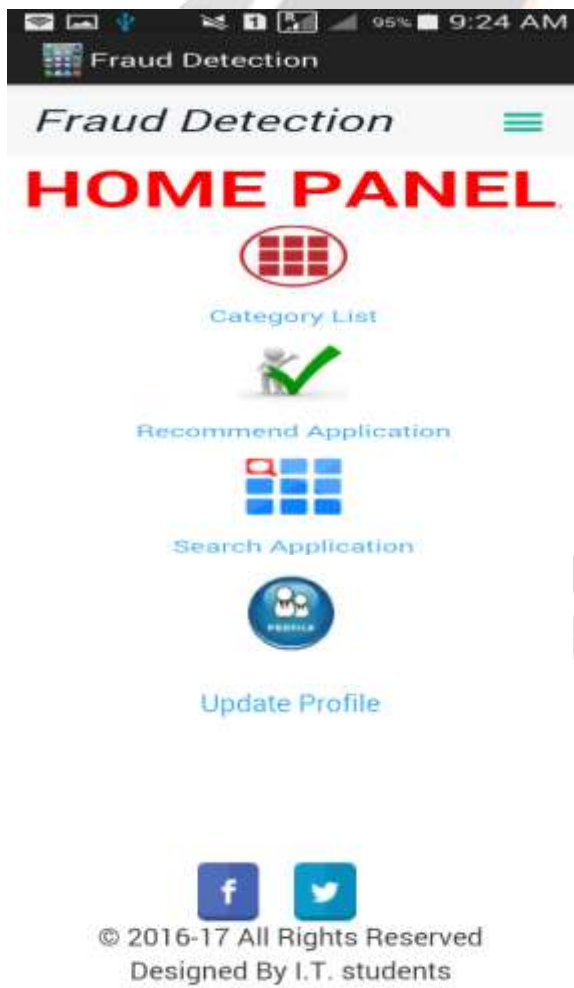
Employed

abc

Submit

Clear

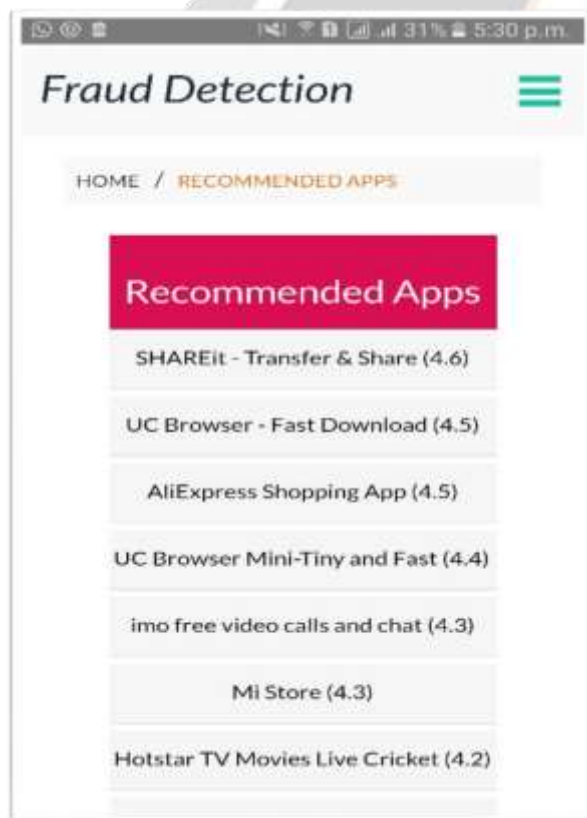
6.1.3 Home Panel



6.1.4 App Category List



6.1.5 Recommendation of apps to user based on age, profession and gender of user.



6.2 Performance Evaluation

For performance evaluation, we have tested our system for 20 apps. Time required for processing is captured for Rank, Rating and Review processing so the total processing time is calculated from it.

Number Of Apps	Rank and rating Processing Time	Review Processing Time	Total Processing Time
5	0.488	2.56	3.048
10	0.915	5.293	6.208
15	1.393	10.145	11.538
20	1.672	16.027	17.699

Table -1

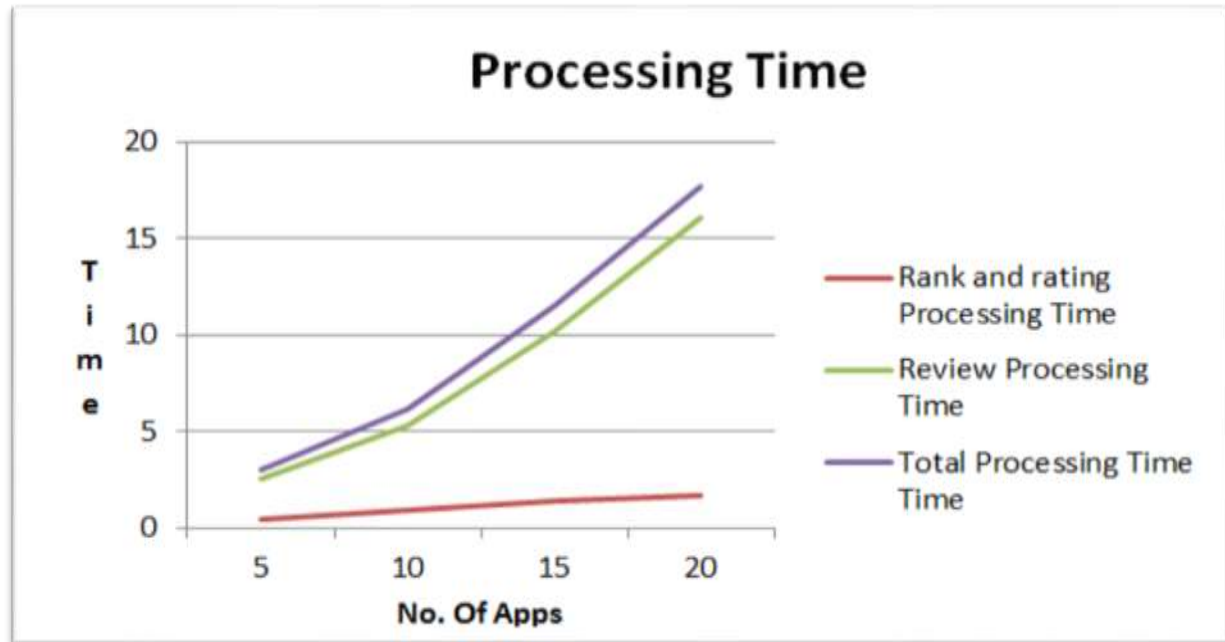


Chart -1: Graphical Representation

7. CONCLUSIONS

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, 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, thus it is easy to be extended with other evidences from domain knowledge to detect ranking fraud.

Finally, we have provided more effective fraud evidences and analysed the latent relationship among rating, review and rankings. Moreover, we have extended our ranking fraud detection approach with other mobile App related services, such as mobile Apps recommendation, for enhancing user experience. We have implemented app recommendation system which recommends in two ways: First according to the age, profession and gender it notifies user which apps the user should use. Second notification is given to the user about the apps which user has not used for longer in his mobile and current rates and reviews for that app.

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