# Firewall Testing For SQL Injection Attack Using Random Forest

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Abstract- Firewalls plays an important role in protecting online system. Existing Firewalls easily detects viruses and malware which are External Files. Firewall check Viruses and Malware from their libraries and then detects and eliminates them. But in SQL Injection, only input is potentially harmful component which restrict the system.

Firewalls cannot parse on run time input. It does not have any malicious file which makes it complex to detect. So, to detect SQL Injection attack, we create virtual Node Which Evaluate Input String and then decides whether input is safe or not. If SQL Injection attack is Detected, It Generates alert and then blocks it, therefore saving the System from Destruction and Unauthorized access to data.

Keywords- machine learning, product.

## I. INTRODUCTION

Firewalls plays an important role in protecting online system. Existing Firewalls easily detects viruses and malware which are External Files. Firewall check Viruses and Malware from their libraries and then detects and eliminates them. In SQL Injection, only input is potentially harmful component which restrict the system. Firewalls cannot parse on run time input. It does not have any malicious file which makes it complex to detect. To detect SQL Injection attack, we create virtual Node. Which Evaluate Input String and then decides whether input is safe or not. If SQL Injection attack is Detected, It Generates alert and then blocks it.

Therefore saving the System from Destruction and Unauthorized access to data. The primary aim of any web application firewall is that of preventing SQL Injection attack. If any attack takes place, it has to save user from that attack and prevent his system and data from damage. System can be damaged to a great extent by SQL injection attack and in some case can be permanently damage. Data stolen can be misused by attacker in any form. The attacker can use the stolen data to gain financial benefit or gain any secret information which may have an adverse impact on user. SQL Injection is more advance attack which is not detected by normal firewall. Therefore, it helps user identify various firewalls and to generate new attack. we found that SQL injection attack is hard to identify when attacker is using so hesitated methods but our system will take data set of such kind of sophisticated attack (strategies and inputs) and compare it with run time input which will identity if the input is vulnerable or not. So, by doing this project, we hope to aware our users about various features of SQL Injection and hope they remain safe. Our efforts will have some fruit to bear and hope our efforts will be recognized and appreciated by others. People will atleast know something about SQL injection attack and will have some basic information. Firewalls plays an important role to protect online system. Nowadays, attacks are becoming more and more sophisticated. They are complex in nature and are not easily recognizable. Many people have no idea about SQL injection attack. Some have, but they take it too lightly. Therefore SQL attack must be taken seriously by everyone.

### **II. ORGANIZATION OF PAPER**

Section 1: In this chapter, introduction of the project i.e. what is need, relevance and what is actual project idea. Section 2: In this chapter, we briefly review the related work on mental disorder detection and their different techniques. Chapter 3: Describe open issues.

Chapter 4: in this chapter, conclusion of the project, Future Scope of the project.

## We are going to use following UML Diagrams:-

- 1. Use Case Diagram
- 2. Entity Relationship Diagram
- 3. Activity Diagram
- 4. Class Diagram

## III. RELATED WORK.

In this paper Author develop a unified detection approach named IMIA-HCRF, to progressively discriminate malicious injection behaviors for recommend er systems. First, disturbed data are empirically eliminated by implementing both the construction of association graph and enhancement of dense behaviors, which can be adapted to different attacks. Then, the smooth boundary of co-visitation behaviors is further segmented using higher order potentials, which is Finally leveraged to determine the concerned injection behaviors[1].

In this work, author propose new attacks to recommend-er systems. Attacks exploit fundamental vulnerabilities of recommend-er systems and can spoof a recommend-er system to make recommendations as an attacker desires. Key idea is to inject fake co-visitations to the system[2].

Author improve detection performance from following two aspects. Firstly, extract well-designed features from user profiles based on the statistical properties of the diverse attack models, making hard detection scenarios become easier to perform. Then, refer to the general idea of rescale Boosting (R Boosting) and Ada-boost, then apply a variant of Ada-boost, called the re-scale Ada-boost (RAdaBoost) as detection method based on the extracted features [3].

In this work, author propose GANG, a guilt-byassociation method on directed graphs, to detect fraudulent users in OSNs. GANG is based on a novel pairwise Markov Random Field that we design to capture the unique characteristics of the fraudulent-user-detection problem in directed OSNs. In the basic version of GANG, given a training dataset, we leverage Loopy Belief Propagation (LBP) to estimate the posterior probability distribution for each user and uses it to predict a user's label[4].

In this paper, author examine the detection of shilling attacks in privacy-preserving collaborative filtering systems. Authors utilize four attack-detection methods to filter out fake profiles produced by six well-known shilling attacks on perturbed data. They evaluate these detection methods with respect to their ability to identify bogus profiles. Real databased experiments are performed. Empirical outcomes demonstrate that some of the detection methods are very successful at filtering out fake profiles in privacy-preserving collaborating filtering schemes[5].

Author provide a formulation for learning to attack a recommend-er as a repeated general-sum game between two players, i.e., an adversary and a recommend-er oblivious to the adversary's existence. We consider the challenging case of poison¬ing attacks, which focus on the training phase of the recommend-er model. Author generate adversarial user profiles targeting subsets of users or items, or generally the top-K recommendation quality. Moreover, author ensure that

the adversarial user profiles remain unnoticeable by preserving proximity of the real user rating/interaction distribution to the adversarial fake user distribution. Author offer a wide range of experiments, instantiating the proposed method for the case of the classic popular approach of a low-rank recommend-er, and illustrating the extent of the recommender's vulnerability to a variety of adversarial intents[6].

In this work, author present Prone - a fast, scalable, and effective model, whose single-thread version is 10–400\_ faster than efficient network embedding benchmarks with 20 threads, including LINE, Deep Walk, node2vec, Gare, and HOPE. As a concrete example, the single-thread Prone requires only 29 hours to embed a network of hundreds of millions of nodes while it takes LINE weeks and Deep- Walk months by using 20 threads. To achieve this, Prone first initializes network embedding s efficiently by formulating the task as sparse matrix factorization. The second step of Prone is to enhance the embeddings by propagating them in the spectrally modulated space[7].

In this work, author propose Ian us, a Sybil detection method that leverages account registration information. Ian us aims to catch Sybils immediately after they are registered. First, using a real-world registration data set with labeled Sybils from We Chat (the largest online social network in China), author perform a measurement study to characterize the registration patterns of Sybils and benign users. Author find that Sybils tend to have synchronized and abnormal registration patterns. Second, based on our measurement results, model Sybil detection as a graph inference problem, which allows us to integrate heterogeneous features[8].

In this work, Author propose Sybil SCAR, a new structure based method to perform Sybil detection in OSNs. Sybil SCAR maintains the advantages of existing methods while overcoming their limitations. Specifically, Sybil SCAR is Scalable, Convergent, Accurate, and Robust to label noises. author first propose a framework to unify RW-based and LBPbased methods. Under our framework, these methods can be viewed as iteratively applying a (different) local rule to every user, which propagates label information among a social graph. Second, author design a new local rule, which Sybil SCAR iteratively applies to every user to detect Sybils. We compare Sybil SCAR with a state-of-the art RAW-based method and a state-of-the-art LBS-based method, using both synthetic Sybil's and large-scale social network datasets with real Sybil's[9].

In this work, author propose a novel collective classification framework to address this long-standing challenge. author first formulate learning edge weights as an

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optimization problem, which quantifies the goals about the final reputation scores that aim to achieve. However, it is computationally hard to solve the optimization problem because the final reputation scores depend on the edge weights in a very complex way. To address the computational challenge, author propose to jointly learn the edge weights and propagate the reputation scores, which is essentially an approximate solution to the optimization problem[10].

## **IV. PROPOSED WORK**

First, we propose to enhance profile injection behaviors and co-visitation injection behaviors via the elimination of disturbed data and representation of sparse behaviors, which also provides a possibility for the integrated detection of different injection attack behaviors. Second, we explore attributes of both nodes and edges of behavior association graph, and propose to incorporate unary potential and pairwise potential of higher order conditional random fields for informative representations of rating and covisitation behaviors. Third, we develop a unified detection approach to identify both profile injection attacks and covisitation injection attacks. Additionally, mixed profile injection attacks and mixed co-visitation injection attacks with different cases are implemented.

## **V. SYSTEM ARCHITECTURE**



Fig 1: System Architecture

In the overall system architecture, a WAF is placed in front of the web application that must be protected. Every request that is sent to the web application is examined by the WAF before it reaches the web application. The WAF hands over the request to the web application only if the request complies with the firewall's rule set. Web application firewalls that aim at preventing SQL injection (SQLi) attacks and develop automated testing techniques that generate SQLi attacks bypassing the WAFs and therefore help uncover holes in WAFs. Mod Security is a popular and representative web application firewall, which is configured to protect a web service applications from SQLi attacks. Web Application Firewalls, web applications with high security requirements are commonly protected by web application firewalls. Sql injection vulnerabilities generally allows an attacker to view data that they are not normally able to retrieve.ML-Driven SQLi Attack Generation is a code injection technique, used to attack data driven applications, in which malicious SQL statements are inserted into an entry field for execution. The two lines about how this paper is concerned with the project undertaken: - It focuses on research on web application firewalls and SQL injection attacks. Web Application Firewalls aim at preventing SQL injection attacks and develop automated testing techniques.

### 1. Use Case Diagram with necessary information

Use case diagram is used to show which operations are performed by the user and which operation are performed by the system.



Fig 2: Use Case Diagram

## 2. Activity Diagram with necessary information



Fig 3: Activity Diagram

Activity Diagram shows the active flow of the system. In above diagram the flow of our project is shown actually how the data flow.

## 3. State Diagram with necessary information

In state diagram step by steps is shown



Fig 4: State Diagram

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#### 4. Class Diagram with necessary information



### Fig 5: Class Diagram

## **VI. ALGORITHM**

Step 1 – First, start with the selection of random samples from a given data set.

Step 2 – Next, this algorithm will construct a decision tree for every sample. Then it will get the prediction result from every decision tree.

Step 3 - In this step, voting will be performed for every predicted result.

Step 4 – At last, select the most voted prediction result as the final prediction result.

## VII. ADVANTAGES

- N To enhance dense rating behaviors and co-visitation injection behaviors via the elimination of disturbed data and representation of sparse behaviors, which also provides a possibility for the integrated detection of different injection attack behaviors.
- $\tilde{\mathbb{N}}$  To develop a novel detection approach to identify both profile injection attacks and co-visitation injection attacks.
- $\tilde{N}$  To display only trusted reviews to the users.
- $\tilde{N}$  To identify spam and spammers as well as different type of analysis on this topic.

## VIII. CONCLUSION

This work presents a machine learning strategy to detect profile injection attacks and co-visitation injection

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attacks for online recommend er systems. Experimental results on both synthetic data and real-world data show that the elimination of disturbed data, determination of disturbed of dense behaviors, and potential segmentation exhibit considerable stability and discriminability among nodes (users or items) for detecting malicious injection behaviors.

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