

A Pragmatic Optimal Approach for Detection of Cyber Attacks using Genetic Programming

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Problem Statement:



- Attackers have been evolving their strategies and methods with time. Using ML/DL methods will certainly improve their exploitations. Therefore, developing a model which fights against such threats is very essential.
- Solving a complex attack has become a very difficult task for security engineers. Hence, building a self learning model will enhance the ability to identify and rectify such attacks.
- Attacks happen when the systems are most vulnerable, such scenarios will be guarded by the developed model.



Statistics:



- **94%** of malware is delivered via mail.
- Phishing attacks account for more than **80%** of reported security incidents.
- **\$17,700** is lost every minute due to phishing attacks
- **60%** of breaches involved vulnerabilities for which a patch was available but not applied
- **63%** of companies said their data was potentially compromised within the last 12 months due to hardware-or silicon-level security breach.
- Data breaches cost enterprises an average of **\$3.92 million**.
- 40% of IT leaders say cybersecurity jobs are the most difficult to fill.



Amount of monetary damage caused by reported cyber crime to the IC3 from 2001 to 2019 (in million U.S. dollars)

Sources

FBI; IC3; US Department of Justice © Statista 2020 Additional Information:

Worldwide; IC3; 2001 to 2019, excluding 2010; Cybercrime reported to IC3





"Intrusion detection is the

process of monitoring the events occurring in a computer system or network and analysing them for signs of possible incidents, which are violations or imminent threats of violation of computer security policies, acceptable use policies, or standard security practices"

• An Intrusion detection system (IDS) is a software application or hardware appliance that monitors traffic moving on networks and through systems to search for suspicious activity and known threats, sending up alerts when it finds such items.







Machine learning/Deep learning is adopted in a wide range of domains where it shows its superiority over traditional rule-based algorithms. These methods are being integrated in cyber detection systems with the goal of supporting or even replacing the first level of security analysts.

The use of a branch of ML namely **Evolutionary Computation (EC)** plays an important role in tracking, analyzing, identifying digital security threats to combat viruses and hackers



Genetic Programming

Executional steps for general GP:

1. Generate an initial population of random compositions (Computer Programs).

2. Run a tournament, which picks four programs randomly out of the population of programs.

3. Apply the search operators crossover and mutation (and possibly others) to the winners

- a. Copy the two winners
- b. With **Crossover** Frequency, apply crossover to copies of the winners
- c. With Mutation Frequency, mutate the programs from (a)
- **4.** Replace the tournament losers with the new offspring.

5. Repeat until a predefined termination criterion has been satisfied, or a fixed number of generations have been explored.

6. The solution is the genetic program with the best fitness within all the generations.









Proposed Method:

- 1) Data Acquisition
- Modern DDoS Dataset is used for implementation.
- A novel Dataset which that contains modern kinds of DDoS attacks.
- Generated using NS2 Network simulator.
- The dataset had 2,160,668 number of instances.

| Variable | Attribute Name | Description |
|----------|-------------------|------------------------------------|
| Number | | |
| 1 | SRC_ADD | Source Address |
| 2 | DES_ADD | Destination Address |
| 3 | PKT_ID | Packet Identifier |
| 4 | FROM_NODE | Source Node |
| 5 | TO_NODE | Destination Node |
| 6 | PKT_TYPE | Packet Type |
| 7 | PKT_SIZE | Total packet size in bytes |
| 8 | FLAGS | Flags |
| 9 | FID | Flow identifier |
| 10 | SEQ NUMBER | Sequence number |
| 11 | NUMBER_OF_PACKET | Total number of packets |
| 12 | NUMBER_OF_BYTE | Total number of bytes |
| 13 | NODE_NAME_FROM | Node Name From |
| 14 | NODE_NAME_TO | Node Name To |
| 15 | PKT_IN | Total time of packet inside queue |
| 16 | PKT_OUT | Total time of packet outside queue |
| 17 | PKT_R | Time of packet received |
| 18 | PKT_DELAY_NODE | Time packet delay within node |
| 19 | PKT_RATE | Average packet rate |
| 20 | BYTE_RATE | Average byte rate |
| 21 | PKT_AVG_SIZE | Average packet size |
| 22 | UTILIZATION | Bandwidth utilization |
| 23 | PKT_DELAY | Total time packet delay |
| 24 | PKT SEND TIME | Time of sending packet |
| 25 | PKT RESERVED TIME | Time of receiving packet |
| 26 | FIRST PKT SENT | Time of first packet sent |
| 27 | LAST PKT RESERVED | Time of last packet received |



Dataset classes

- Smurf: The target server receives huge number of ICMP echo requests packet.
- **UDP Flood**: A massive amount of UDP traffic is sent to inundate the server.
- **SQL Injection DDOS**: Sql sentences are used to flood the server.



- **HTTP Flood**: attacker overwhelm the server using HTTP GET/POST methods.
- Normal transaction data.

| Category | No. of Records | |
|------------|----------------|--|
| Smurf | 12,590 | |
| UDP Flood | 201,344 | |
| SIDDOS | 6,665 | |
| HTTP Flood | 4,110 | |
| Normal | 1,935,959 | |



2) Preprocessing

- **Principal Component Analysis (PCA)** was applied on the dataset.
- The features were reduced to 8, 16 and 20 principal components.
- There was no significant difference in the 16 and 20 principal components in terms of percentage loss.
- Hence for further processing, 16 principal components were considered for simplicity.

| Number of Principal Components | % Information Gain | % Information Loss |
|--------------------------------------|-----------------------|-----------------------|
| 8 | 94.92% | 5.08% |
| 16 | 98.48% | 1.52% |
| 20 | 99.6% | 0.4% |



3) Implementation of Genetic Programming:



- For implementation, Distributed Evolutionary Algorithms in Python (DEAP) and Tree-Based Pipeline Optimization Tool (TPOT) frameworks were used.
- Performed in 4 steps:
 - **1.** Build an appropriate type of problem.
 - **2.** Creating a fitness class using Creator module.
 - **3.** Initialization of operators using toolbox operator.
 - **4.** Constructing the main function.





Experiments and Results





- The experiments were performed on six Machine Learning models.
- Values were evaluated using Confusion matrix.

| Classification Method Used | Correct Classification % Score | |
|--------------------------------|--------------------------------------|--|
| KNN | 98.57% | |
| Naive Bayers | 96.66% | |
| Logistic Regression | 98.62% | |
| Decision Tree | 97.38% | |
| Random Forest | 98.02% | |
| Stochastic Gradient Descent | 98.55% | |



GP Results

- GP implementation depends on various parameters such as population size, no. of generations, crossover and mutation rate, etc.
- When the population size was considered as 50 with a crossover rate of 0.01, an accuracy of 98.67% was obtained.
- It was observed that as the no. of generations were increased, the accuracy did not change much.



Comparative Study



| Author | Model | Dataset | Accuracy |
|--|----------------------------------|---------------------------|------------------|
| Hasanen Alyasiri John Clark and Daniel Kudenko | Cartesian Genetic Programming | Modern DDoS | 97.19% |
| Manjula Suresh and R. Anitha | Naive Bayes | CAIDA | 97.2% |
| Mouhammd Alkasassbeh, Ahmad B.A Hassanat, Ghazi Al-Naymat, Mohammad Almseidin | Random Forest Naive Bayes | Modern DDoS | 98.02% 96.91% |
| Naveen Bindra, Manu Sood | Random Forest | CIC IDS 2017 | 96.13% |
| S. Umarani, D. Sharmila | Naives Bayes | 1998 World Cup Website | 95.95% |
| Proposed Model | Genetic Programming | Modern DDoS | 98.67% |

Conclusion



- This scientific analysis investigates an application of Genetic Programming (GP) for intrusion detection. For this study, the Modern DDoS dataset is used. This dataset contains contemporary threats gathered from various environments.
- The proposed GP model detects DDoS attacks with improved accuracy of **98.67%** while comparing it with six established classification models. The obtained results highlight the advantages of adopting the GP model.
- However, it was observed that adopting other approaches for operations such as mutation or crossover can result in better results. Due to limited resources, this was not tested.
- In future, this model can be investigated for other types of attacks and also to come up with a universal model to detect all kinds of well-known threats.