

Intrusion Detection with Genetic Algorithms and Fuzzy Logic

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The Big Picture

- Computer lab gets large numbers of login attempts that are attempts at intrusion.
- An attack could be trying to gain root access to the system.
- With an intrusion detection system (IDS), it would be possible to classify those attempts into legitimate and illegitimate attempts to login.
- Intrusion detection systems provide one way of detecting attacks by monitoring network activities for malicious or abnormal behaviors and then producing reports, alerts, and actions.
- A way of training an IDS about possible threats is by using a fuzzy genetic algorithm.

Outline

- 1 Background
- 2 Using Fuzzy Genetic Algorithms
- 3 Conclusions

Outline

- 1 Background
 - Types of Networking Attacks
 - Detection Methodologies and Rules
 - Data Sets - KDD99 and RLD09
 - Determining the Accuracy of an Algorithm
 - Genetic Algorithms

2 Using Fuzzy Genetic Algorithms

3 Conclusions

Types of Networking Attacks

- Denial of Service (DoS): attacker makes a machine inaccessible to a user by making it too busy to serve legitimate requests.
 - Many systems lock out a user from an account after a certain number of failed login attempts. An attacker would be able to use this to prevent legitimate users from logging in, by intentionally failing to log in enough times to lock the account.
- Probe: attacker examines a machine in order to collect information about weaknesses or vulnerabilities that in the future could be used to compromise the system.
 - An attack could be trying to determine what version of a software is being run on that machine, and if that version has a known issue then that allows them to attempt to attack that.

Detection Methodologies

- Signature-based detection: compares well-known patterns of attacks that are already known to the IDS against captured events in order to identify a possible attack.
 - Simple and effective way to detect known attacks, but is ineffective against new kinds of unknown attacks.
- Anomaly-based detection: looks for patterns of activity that are rare and uncommon.
 - Harder to do than signature-based detection, but it can be an effective way to detect new, unknown attacks.

Rules

- A commonly used approach for detecting intrusions is to use rules.
- If-Then format: If (*condition*) then (*consequence*).
 - The condition is composed of one or more features, and the consequence says if it is an intrusion or not.
 - If *duration* = 4 then *intrusion*.

KDD99

- Generated by simulating a military network environment in 1999.
- Has long been a standard data set for intrusion detection.
- Data was processed into 5 million *records*.
 - A record is a sequence of TCP packets, between which data flows to and from a source IP address to a target IP address.
- Each record is classified as either normal or attack activity.
- KDD99 uses 41 *features*, which are properties of a record that are used to describe the activity and help to distinguish normal connections from attacks.

Some Features of KDD99

- duration: length of the record in seconds.
- num_failed_logins: number of failed login attempts.
- root_shell: returns 1 if root shell is obtained, else returns 0.

RLD09

- RLD09 was created because KDD99 is 14 years old, and newer attack types are not included in KDD99 because of its age.
- Data was captured from a university in Bangkok, Thailand.
- As well as normal network activity, RLD09 has 17 different types of attacks (divided into denial of service and probe attacks).

Training and Testing Sets

- In a machine learning experiment, a common technique is to divide the data set into two subsets, a *training set* and a *testing set*.
- The given algorithm is then trained on the training set to look for patterns.
- These patterns are then verified using the test set.

Determining the Accuracy of an Algorithm

		Predicted	
		Not Attack	Attack
Actual	Not Attack	True Negative (TN)	False Positive (FP)
	Attack	False Negative (FN)	True Positive (FP)

- Detection rate (DR): the percentage of normal and attack activity correctly classified from the total number of data records.

Genetic Algorithms

- GAs are a search technique used to find solutions to problems.
- Possible solutions to problems can be represented in a variety of problem dependent ways, such as bit strings.
 - IDS rules can be represented as bit strings.
- First, a randomly generated population of potential solutions is created. Then mutation, crossover, and selection are applied to each generation until an acceptable solution is found or some time limit is exceeded.

Genetic Algorithms

- Mutation: random bits in an individual, or possible solution, are randomly changed.
- Crossover: two individuals swap sequences of bits to form two new individuals.
- In an IDS: mutation takes the bits of a rule and changes them to form a slightly different rule. Crossover takes 2 rules and creates new rules by swapping the bits of the old rules.

Genetic Algorithms

- Selection: individuals that have better fitness are chosen to be parents.
- The fitness of an individual is specified by the fitness function, which determines the quality of a particular individual.
- In an IDS: fitness measures how well a rule classifies records as either attacks or normal activity. Selection combined with a fitness function directs the search towards an effective solution.

Outline

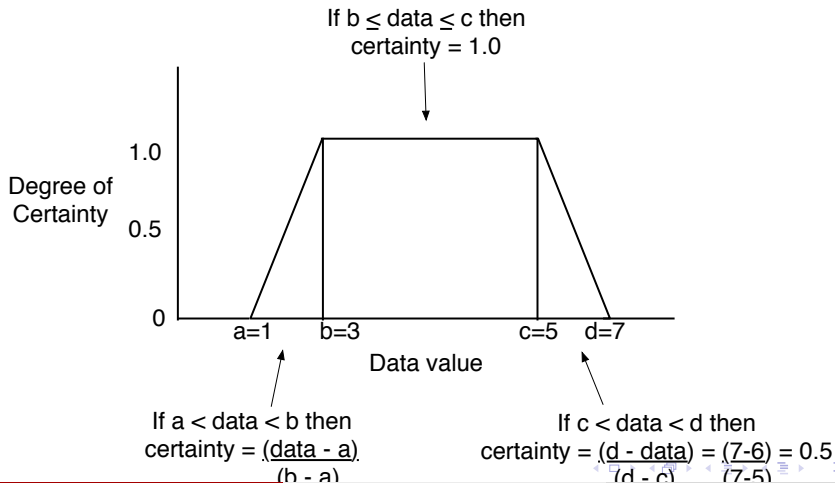
- 1 Background
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 - Fuzzy Algorithm
 - Algorithm Overview
 - Experimental Design and Results
- 3 Conclusions

Fuzzy Logic

- Attacks on systems do not always have a fixed pattern, so fuzzy logic is used to detect patterns that have a behavior that is between normal and unusual.
- Fuzzy logic rules are similar to the rules described before, except that consequence is a certainty factor.
 - If (*duration* = 6) then (*the degree of certainty of the record being an attack is 0.5*).

Finding the Degree of Certainty of a Record Being an Attack

Suppose that the feature is duration, and it is 6 seconds. Then data=6.



Encoding of Features and Rules

- The four parameters are encoded into blocks. Each block is a feature with values between 0 and 7.

010	011	100	101
a=2	b=3	c=4	d=5

- A rule has one block for each of 12 features followed at the end by a marker indicating the type of attack.

010	011	100	101	010	011	101	111	DoS
a=2	b=3	c=4	d=5	a=2	b=3	c=5	d=7	
Block 1					Block 12				Type

Figure : Based on [Jongsuebsuk-Wattanapongsakorn-Charnsripinyo, 2013]

- The degree of certainty is computed for each of the 12 blocks, and if the sum of those is greater than a threshold, then it will be declared as an attack.

Algorithm Overview

- One record is passed into a rule.
- Each feature in a record is matched to one block of the rule.
- The parameters of each block measure the degree of certainty of an attack using the trapezoidal fuzzy rule shape.
- The sum of the degrees of certainty from each block are then compared with a threshold to determine if the record represents an attack or normal behavior.

Algorithm

The fitness function in the algorithm is:

$$\frac{\alpha}{A} - \frac{\beta}{B}$$

α : # of attack records correctly identified as attack.

A : # of attack records.

β : # of normal records incorrectly classified as attack.

B : # of normal records.

Experiments

- A variety of experiments were run. Two experiments used just RLD09. Three experiments used both RLD09 and KDD99 in order to compare how the fuzzy GA would perform on both.

Experiments Using Only RLD09

Experiment 1

- Fuzzy GA was used to create DoS and probe detection rules. Both the DoS and probe rules were then used together in the testing process to identify attacks from the testing data set.
- 10,000 records in the training set, 26,500 records in the test set.
- The following algorithm was used to identify attacks and normal activity in this experiment.

```
if dos_rule = yes or probe_rule = yes then  
    This record is an attack;  
else  
    This record is normal;  
end if
```

Experiments Using Only RLD09

Experiment 1 Results

	Attack	Normal	Total	FP(%)	FN(%)	DR(%)
DoS Training	1499	8501	10000	1.46	47.50	91.64
Probe Training	2496	7504	10000	1.83	15.38	94.79
Testing	10500	16000	26500	1.13	4.10	97.92

Experiments Using Only RLD09

Experiment 2

- Attacks were pulled out of the training set and kept for unknown data testing. This was to test that the fuzzy GA could detect unknown attacks.
- Used fuzzy GA and a decision tree algorithm, which is another common algorithm for classification problems.
- 7 tests were run, and for each test case there were 13 attack types plus normal activity that were in the training data set. Three attack types were used for the unknown testing data set.

Experiments Using Only RLD09

Experiment 2 Results (7 tests were run in total, 5 are shown here.)

Test Case	Unknown Attacks	Decision Tree DR (%)	Fuzzy Genetic DR (%)
1	Adv Port Scan (Probe)	Avg =	Avg =
	Ack Scan (Probe)	98.33	100
	Xmas Tree (Probe)		
2	UDP Flood (DoS)	Avg =	Avg =
	Host Scan (Probe)	46.65	99.80
	UDP Scan (Probe)		
3	Jping (DoS)	Avg =	Avg =
	Syn Scan (Probe)	99.70	98.75
	Fin Scan (Probe)		
4	UDP Flood (DoS)	Avg =	Avg =
	RCP Scan (Probe)	70.35	98.15
	Fin Scan (Probe)		
5	Http Flood (DoS)	Avg =	Avg =
	RCP Scan (Probe)	99.94	97.50
	Fin Scan (Probe)		

Experiments Using Both RLD09 and KDD99

Three experiments used both RLD09 and KDD99.

Experiment 1

- Used fuzzy GA to classify normal activity and attacks from KDD99 and RLD09.

Data set	Attack	Normal	FP (%)	FN (%)	DR (%)
KDD99	160,117	39,337	0.13	1.55	98.72
RLD09	10,500	16,000	1.14	3.39	97.97

Experiments Using Both RLD09 and KDD99

Experiment 2

- Used the fuzzy GA to classify types of attacks in KDD99.
- 158,597 records of DoS attacks. 1,500 records of probe attacks.
- 10 tests were run in total, 5 are shown here.

Test	Attack	Type	FP (%)	FN (%)	DR (%)
1	Back	DoS	85.33	0.00	16.56
2	PoD	DoS	84.66	0.00	15.58
3	Smurf	DoS	0.76	0.10	99.73
4	PortswEEP	Probe	6.40	0.00	93.66
5	Satan	Probe	0.74	3.75	99.22

- 8 test cases had DR greater than 93%. Only 2 cases had low DR.

Experiments Using Both RLD09 and KDD99

Experiment 3

- Used the fuzzy GA to classify types of attacks in RLD09.
- 6,400 records of DoS attacks. 10,400 records of probe attacks.
- 17 tests were run in total, 6 are shown here.

Test	Attack	Type	FP (%)	FN (%)	DR (%)
1	HTTP Flood	DoS	0.36	3.5	99.46
2	Smurf	DoS	0.02	0	99.98
3	UDP Flood	DoS	11.06	0	89.59
4	Fin Scan	Probe	2.58	0	97.50
5	IP Scan	Probe	13.01	16.4	86.89
6	Syn Scan	Probe	0.65	4.2	99.24

- 15 cases had DR greater than 97%. 2 cases had low DR, (cases 3 and 5).

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Conclusions

- The fuzzy genetic algorithm had a higher detection rate than a decision tree algorithm in most cases.
- Fuzzy genetic algorithms are good at detecting unknown attacks.
- The use of fuzzy genetic algorithms in intrusion detection is an effective way of detecting attacks.

Thanks!

Thank you for your time and attention!

Questions?

References



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See my Senior Seminar paper for additional references.