#### Fast and Scalable Method for Resolving Anomalies in Firewall Policies

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## Introduction

- Firewalls
- Types of Anomalies
- Related Work
- Data Structure and Algorithm
- Experimental Results
- Conclusion

## **Firewalls**

#### Firewall

- System acting as an interface of a network to one or more external networks.
- Implements the security policy of the network
  - By deciding which packets to let through
    - Based on rules defined by the network administrator.



## Example

Id	Protocol	Source		Destination		Action	Probability
		IP address	Port	IP address	Port		
$r_1$	TCP	71,123,10,*	any	10.0.0.1	21	pemit	0.3
$r_2$	TCP	*,*,*,*	any	10.0.0.1	21-23	deny	0.25
$r_3$	TCP	71.*.*.*	any	10.0.0.1	21	pemit	0.15
$r_4$	TCP	71.123.*.*	any	10.0.0.1	21	deny	0.1
$r_5$	TCP	* * * *	any	10.0.0.1	23-25	pemit	0.1

### **Protection Methods**

- Firewalls Firewall policy rules should be designed carefully!
- Challenges
  - Rules are created by multiple people
  - Rules are created over extended period of time
  - Number of rules in a firewall policy can be 5K+!
  - Rules are dynamic!

#### Relationships Between Rules -Disjoint Rules



Two rules r and s are disjoint if they have at least one criterion for which they have completely disjoint values

#### Relationships Between Rules -Exactly Matching



Two rules r and s are exactly matched if each criterion of the rules match exactly.

#### Relationships Between Rules -Inclusively Matching (Shadowing)



Two rules r is a subset, or inclusively matched of another rule s if there exists at least one criterion for which r's value is a subset of s's value and for the rest of the attributes r's value is equal to s's values.

#### Example:

IN, TCP, 64.233.179.104, 80, 192.168.20.3, ANY, ACCEPT>
IN, TCP, 64.233.179.104, ANY, 192.168.20.\*, ANY, ACCEPT>

#### Relationships Between Rules -Correlated



Two rules r and s are correlated if r and s are not disjoint, but neither is the subset of the other.

## **Existing Work**

- E. W. Fulp O(n^3) algorithm to order rules in a given policy; it doesn't discover correlated ones.
- E. Al-Saher *et al.* Method for selecting rules based on their probability.
- A. Liu Method to discover and remove redundant rules (Exact matching).

## **Our Approach**

- We aim at removing few troublesome rules from given policy to resolve anomalies.
- Design a data structure to represent dependencies among rules.
- Remove troublesome rules.
- Return a subset of consistent rules and correlated rules (for editing).

## **Our Approach**

- Design a data structure to represent dependencies among rules.
- Graph D is directed, and U is undirected.
  - Each node in U represents a rule
  - Two nodes are connected in U if there is shadowing or correlation relationship between these two rules.
- Graph D describes dependency among rules.

## **Our Approach**

- Select a rule that doesn't depend on any other rule (terminal node) from D.
- Remove corresponding links from U and links/nodes from D.
- If graph U is disconnected and new component formed, continue, else there is correlation
- If there is correlation, choose the rule with highest probability.

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### **Example – Our Approach**



#### Complexity

O(n^2) to construct graphs *D* and *U* O(2log *n*) to discover dependencies
 Algorithm complexity O(n^2 log n)

#### **Experimental Results**

- Two sets of test experiments executed:
  - Real-life tests: five policies of size 107, 361, 647, 881, and 1385 over a month period on Verizon firewall using the original (non-improved) approach.
  - Tests done over the same period using improved approach.
- Five test sets have been executed on synthetic policies of sizes 10K 30K.

# Experimental Results – Real-Life Policies

Test	No. of	Avg. Base	Avg.	Avg.	Imp.
	Rules	Comp.	Correlation	Dependency	Ratio
1	107	43.1	2.7	8.5	63.3%
2	361	87.2	1.4	2.2	47.2%
3	647	381.1	3.1	7.9	62.7%
4	881	341.6	3.3	6.4	71.2%
5	1385	715.3	3.8	6.7	74.8%

# Experimental Results – Synthetic Policies

Test	No. of	Avg. Base	Avg.	Avg.	Imp.
	Rules	Comp.	Correlation	Dependency	Ratio
1	10K	4224	121.3	13.5	68.6 %
2	12.5K	5584	389.6	11.6	40.5 %
3	15K	8054	274.7	12.0	76.4 %
4	25.5K	14263	649.2	15.2	79.3 %
5	30K	17714	712.4	20.7	87.6 %

### **Current & Future Work**

- Find exact minimum number of rules to eliminate all anomalies from policy.
- Modify algorithm to handle dynamicpolicies.
- Improve the algorithm performance.

### **Thank You All!**

**Questions?**