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An Optimized Approach for Assisted Firewall Anomaly Resolution

DANIELE BRINGHENTI1, LUCIA SENO2, and FULVIO VALENZA1

1 Dipartimento di Automatica e Informatica, Politecnico di Torino, Torino, Italy (e-mail: daniele.bringhenti@polito.it, fulvio.valenza@polito.it)
2 Cnr Istituto di Elettronica e di Ingegneria dell’Informazione e delle Telecomunicazioni (CNR-IEIIT), Padova, Italy (e-mail: lucia.seno@cnr.it)

Corresponding author: Daniele Bringhenti (e-mail: daniele.bringhenti@polito.it).

ABSTRACT The security configuration of firewalls is a complex task that is commonly performed manually by network administrators. As a consequence, among the rules composing firewall policies, they often introduce anomalies, which can be classified into sub-optimizations and conflicts, and which must be solved to allow the expected firewall behavior. The severity of this problem has been recently exacerbated by the increasing size and heterogeneity of next-generation computer networks. In this context, a main research challenge is the definition of approaches that may help the administrators in identifying and resolving the anomalies afflicting the policies they write. However, the strategies proposed in literature are fully automated, and thus potentially dangerous because the error-fixing process is not under human control. Therefore, this paper proposes an optimized approach to provide assisted firewall anomaly detection and resolution. This approach solves automatically only sub-optimizations, while it interacts with human users through explicit queries related to the resolution of conflicts, as their automatic resolution may lead to undesired configurations. The proposed approach also reduces the number of required interactions, with the aim to reduce the workload required to administrators, and employs satisfiability checking techniques to provide a correct-by-construction result. A framework implementing this methodology has been finally evaluated in use cases showcasing its applicability and optimization.

INDEX TERMS firewall, policy anomaly management, policy-based systems

I. INTRODUCTION

In modern computer networks, firewalls are pervasively deployed to protect network assets from cyber threats by blocking undesired and malicious traversing traffic [1]. Firewalls can be either dedicated hardware devices or software programs running on general-purpose servers, relying on traditional operating systems or virtualized environments, as in the case of networks based on the Network Functions Virtualization (NFV) paradigm. Firewalls are commonly placed at the edge of the protected network, to act as a first line of defense against attacks coming from the outside. Also, increasingly often, firewalls are deployed within the protected network, following the defense in depth principle. The latter design allows to differentiate among globally and locally enforced security policies, enables fine-grained control of traffic flowing within the network and fast attack response, as well as limits the propagation of successful both outside and inside attacks.

The way a firewall filters traversing traffic is dictated by its configuration, establishing which packets must be forwarded on the next hop toward their destination, and which ones must be discarded as undesired or malicious. Writing and maintaining firewall configurations is a cumbersome task that is still often performed manually by network administrators, i.e., without the support of automatic or semi-automatic tools. As a consequence, the probability of anomalies (i.e., sub-optimizations and conflicts) affecting firewall configurations is considerably high, especially in the case of complex, heterogeneous networks, characterized by the presence of multiple firewalls with a high number of rules. Firewall misconfigurations, in particular, have proven to recurrently lead to unfortunate events such as data breaches or access control violations. According to a recent Data Breach Investigations Report produced by Verizon, in 2021 82% of breaches were caused by human errors in security configuration [2]. The Verizon analysis clearly shows that traditional manual approaches to firewall configuration reveal inefficient and unreliable and are not the best course of action for modern networks.

For the above reasons, one of the research challenges that...
has been addressed in the last decade is the development of methodologies and tools for identifying and resolving anomalies afflicting human-written firewall configurations. This problem is quite complex, as multiple anomaly classes must be accounted for, and those highlighting policy sub-optimizations and errors must be detected and fixed. Some state-of-the-art methodologies have been proposed to perform both the detection and resolution tasks automatically. However, fully automated anomaly resolution strategies are potentially very dangerous as the error-fixing process is not under the network administrator control. As an example, if multiple conflicting firewall rules are detected, a fully automated tool may make a decision that is not coherent with the original vision, i.e., the security intents of administrators.

In light of these considerations, this paper proposes an optimized approach to provide assisted resolution of intra-firewall policy anomalies (i.e., anomalies that occur within the same firewall rule list). This approach automates all anomaly resolution operations that do not require external human interventions, i.e., the ones related to sub-optimizations. Instead, it interacts with network administrators through explicit queries related to the resolution of conflicts, which, if automatically addressed, may lead to undesired resulting configurations. The approach is able to remove all rules that are unnecessary in the firewall policy and to reorder the remaining rules to create a final anomaly-free firewall policy. The scope of the proposal lies in intra-firewall policy resolution, because addressing the issue of solving all anomaly types for a rule list is a research problem complex by itself, as also proved by the large number of studies proposed in literature within the same scope.

The main novelties of this proposal are the following ones:

- **Differently from state-of-the-art approaches, which are still manual or fully automatic, our proposed approach is semi-automatic. This feature contributes to avoid that the human user manually solves each anomaly, but at the same time it allows him to interact with the fixing process to define his operational guidelines, thus overcoming the previously mentioned limitations of the related literature.**

- **The proposed approach aims to reduce the workload required to administrators. In fact, the proposed methodology has been designed so as to reduce the number of queries that are actually needed to solve all errors in the original firewall configuration. This feature makes the process swifter and suitable for large firewall configurations where the number of detected anomalies may be consistently high.**

- **The ordering of the firewall rules composing the final anomaly-free firewall policy is established by employing satisfiability checking, and modeling the problem as a Satisfiability Modulo Theories (SMT) problem. Employing an automated SMT solver, whose algorithms have already been proved correct and sound, provides assurance about the outcome of this final operation, which can be said to be correct by construction. Moreover, as reported in the literature [3], the output of this mathematical formulation does not require the application of any other a-posteriori formal verification, and state-of-the-art open-source SMT solvers can efficiently solve SMT problem instances in polynomial times on average.**

The remainder of this paper is structured as follows. Section II discusses related work, highlighting the differences with respect to our proposal. Section III provides an overview of the proposed approach for assisted firewall anomaly detection and resolution. Section IV describes the firewall and rule relationship models, which are at the basis of the proposed methodology. Section V discusses the strategy used for the analysis of firewall policy anomalies. Section VI provides a detailed description of the designed algorithm for assisted anomaly resolution. Section VII describes how the algorithm has been implemented in a prototype tool and validated. Finally, Section VIII draws conclusions and discusses future work.

### II. RELATED WORK

In the latest years, policy-based management has again become a relevant network management paradigm. According to the original definition provided in the RFC-3198 (Terminology for Policy-Based Management) [4], “a network security policy is a set of rules to administer, manage, and control access to the network resource”. Specifically, when a network security policy is a firewall policy (i.e., it expresses information about a firewall behavior), each rule composing it typically has the following structure:

```
< order >< src_IP >< src_Port >< dst_IP >
< dst_Port >< protocol >< action >
```

The filtering rule order determines its position relative to other filtering rules. The IP addresses can be the address of a specific network peer (e.g., 140.192.37.120) or a network address (e.g., 140.192.37.*). Ports can be either a specific port number, a range, or a placeholder, indicated by ‘any’ or ‘*’, to represent all port numbers. A filtering action can be ‘allow’ or ‘deny’.

As recently shown in the survey by Jabal et al. [5], several studies have been proposed in the literature on a specific operation related to network security policies, i.e., policy anomaly analysis. This operation aims to check three main properties of a set of network security policies: correctness, consistency, and minimality.

- **Correctness analysis consists in identifying errors that may have been introduced in policy specification so that the policies are free of faults and compliant with their intended goals and system requirements. Ensuring policy correctness also includes validating their syntax and verifying their ability to achieve their goals in all possible contexts and scenarios.**

- **Consistency analysis consists in identifying conflicts among the policy rules, i.e., identifying all the situations wherein the presence of a rule impacts the behavior of**
another one. For example, a rule contradicts another one if it is applied to the same packet set, but enforces a different filtering action. 

- Minimality analysis consists in ensuring that the policy rule set does not include redundant rules, i.e., detecting sub-optimizations. Redundant rules increase the administrative work required to manage the policy rules and decrease the overall performance, without providing any benefit.

If the analyzed policy rule set lacks any of these three properties, then policy anomaly resolution is the operation that is executed to remove the identified anomalies and reinstate the missing properties.

Policy anomaly analysis does not correspond to policy verification, even if both operations are key concepts of policy-based management. Differently from anomaly analysis, policy verification consists in checking whether a policy rule set is correctly enforced in a system or, in other words, whether the system implementation actually matches the policy semantics [6]–[9]. Therefore, policy verification is commonly an operation that is performed after policy analysis has checked the policy rule set is correct, consistent and minimal, and after it has been enforced on the actual security system. Besides, policy anomaly analysis is also different from policy refinement [10]–[13], which consists in transforming high-level policies into low-level configurations, and usually is a complementary operation in policy-based management. In fact, sometimes performing a full refinement of security policies is not feasible because it would be time-consuming when applied to large networks.

Policy anomaly analysis and resolution have been investigated for the vast domain of access control policies, which may have different characteristics and features. Some of the most relevant studies about it are [14]–[17]. However, the scope of this work lies in the anomaly management of a specific kind of security policy, that is inter-firewall policy. Therefore, as our proposed approach deals with both firewall policy anomaly analysis and resolution, this section analyses research studies related to those two operations, and it compares them with our proposal. In particular, Subsection II-A reports the most significant contributions concerning firewall policy anomaly analysis. Instead, Subsection II-B describes the most effective tools and approaches proposed for resolving firewall anomalies and the relations and differences with our approach.

A. FIREWALL POLICY ANOMALY ANALYSIS

The study by Al-Shaer and Hamed [18] was the first one in literature to address the anomaly analysis problem for firewall policies. This initial study presents a classification scheme for packet filtering rule relations, defining four types of intra-firewall policy rule anomalies, i.e., anomalies that may occur in the same firewall instance. Later, Al-Shaer et al. [19] extend this initial work to detect inter-firewall policy rule anomalies, i.e., anomalies related to policies belonging to different firewall instances in a distributed firewalling architecture.

In particular, the inter-firewall policy analysis evaluates rule relations between serially-connected packet filters.

Based on these initial studies of Al-Shaer et al., other researchers proposed alternative models and classification schemes. The most significant ones are [20]–[22]. Specifically, FIREMAN [20], proposed by Yuan et al., uses binary decision diagrams to represent packet filtering policies. Golnabi et al. [21] extend Al-Shaer et al.’s analysis using a data mining technique, named association rule mining. The anomaly detection based on the mining exposes many hidden but not detectable anomalies by analyzing only the firewall policy rules. Such approach allows the identification of two new non-systematic misconfiguration anomalies: blocking existing service and allowing traffic to non-existing service anomalies. The first misconfiguration case blocks legitimate traffic from a trusted network to an “existing” service, while the other case of the misconfiguration permits traffic destined for a non-existing service. Then, Bouhoula et al. [22] propose a different approach, based on an inference system to detect intra-firewall policy anomalies. They use the inference system to construct a tree representation of the policy, so that this process stops the construction of a specific branch when no anomaly can be found.

The investigation of the anomaly analysis problem has continued up to now, with another set of more recent studies [23]–[26]. In greater detail, the anomaly analysis modules developed by Togay et al. [23] is based on Prolog, a programming language that adopts the paradigm of logic programming. There, firewall policy rules represent nonconditional information and are transformed into fact clauses, so that anomaly detection predicates can be used to make field-wise analyzes between rule pairs. Lin et al. [24] use an asymmetric double decision tree to detect anomalies. In particular, when detecting an incremental anomaly with respect to a policy, they only need to compare the asymmetric simplified equivalent decision tree to find other anomalies. Both the proposals by Komadina et al. [25] and by Andalina et al. [26] perform anomaly analysis applying data mining and machine learning on data logs. However, machine learning represents a less commonly pursued strategy to address this problem, because the accuracy rate of the results is not 100%, and generally unsupervised methods struggle to detect policy anomalies.

The main limitation of all studies mentioned so far, from Al Shaer et al.’s preliminary work to the most recent ones, is that they exclusively address the firewall anomaly analysis problem. After the strategies proposed in these studies identify an anomaly, they are not removed or solved from the original policy set. Instead, differently from them, our proposed approach also embeds a resolution strategy, which helps human administrators solve the identified anomalies through a reduced number of interactions with them.

Another relevant classes of related work about the topic of firewall anomaly analysis is composed of studies ( [27]–[34]) that propose platforms based on Graphic User Interfaces (GUIs), which provide functionalities such as zooming and manual adjustment of the firewall filtering effect areas to help.
human administrators to identify firewall policy anomalies. Specifically, Tran et al. [27] propose PolicyVis, a 2D graphic interface that allows a thorough visualization of all possible firewall behaviors by considering all rules fields, actions and orders. PolicyVis is flexible enough to let users choose their desired fields as 2D graph coordinates, and eases the anomaly analysis operation introducing the features of compressing, focusing, zooming, and segmentation. Chao [28] creates a systematic visualization for the firewall Rule Anomaly Relation tree (RAR tree) that is created on the basis of the investigated firewall rules. Callouts, arrays and other graphical symbols are used to depict intuitively the presence of anomalies, and a specific GUI view is employed to categorize them. Mansmann et al. [29], [30] introduce a graphical tool named Visual Firewall Editor, which depicts a hierarchical view on firewall rules by using the Sunburst visualization. This tool focuses on two main tasks, i.e., to explore firewall rules and object groups, and to test certain hypotheses about a rule set (e.g., to assess the presence of anomalies). In order to perform them, it uses an interactive tree view to show the firewall rule list, the object group hierarchies and enable navigation in them, with features such as auto-zooming and scaling. Kim et al. [31] propose Firewall Policy Checker to further improve user-friendliness in anomaly analysis. Their tool offers a 3D visualization based on two main graphical views. Their representation is composed of two major views. The first one provides an overall view to the entire firewall policies and the detected anomalies, represented by spheres of different colors. The second one allows to perform risk and illegal service discovery, on the basis of the anomalies identified with the first view. Kim et al. [32] realized the visualization of firewall policies in a 6-dimensional space in 3D format through a tool named FRuVATS. Its visualization model allows to represent all ranges of source and destination IP addresses, thus enabling control on each application service, to identify active and inactive domains, and to detect possible anomalies. Lee et al. [33] propose HSViz, a firewall policy visualization tool that offers multiple view to analyses firewall policies. Among them, the most useful ones are the hierarchy view, which visualizes firewall policy ranges based on destination IP octets at the user’s choice, and the anomaly and distributed views, which represent the policy in parallel coordinate charts, easing anomaly detection. Kim et al. [34] introduces a system, named F/WVis, that allows to visualizes both firewall policies and anomalies on the layout of an user interface with a 3D representation. From that, the user can almost immediately identify how many rules composed the firewall policy and which rule has anomalies with other rules from the same policy. Its GUI also eases the investigation of the firewall policy with a searching tool.

Even if these GUI-based approach can help humans in detecting anomalies, the resolution must still be performed manually. Therefore, those manual human operations are still prone to errors, and they may introduce even a higher number of anomalies. Differently from those GUI-based approaches, our proposal introduces a human-assisted semi-automatic mechanism, where the resolution is guided by human answers, but without requiring the users to actually perform it. This guarantees higher efficiency and avoidance of errors, also thanks to the SMT formulation relying on the correctness-by-construction principle.

B. FIREWALL POLICY ANOMALY RESOLUTION

A second class of studies ([35]–[40]) proposes algorithm for anomaly resolution. However, all of them have shortcomings that our approach aims to overcome.

Liu et al. [35] and Jeffrey et al. [40] propose resolution techniques that simply focus only on detecting and removing redundant rules. On the one hand, Liu et al. [35] categorize redundant rules into two main categories: upward redundant rules and downward redundant rules. Upward redundant rules are rules that are never matched, whereas downward redundant rules are rules that are matched but enforce the same action as rules with lower priority. These sub-optimizations are detected and solved through a model based on a data structure named Firewall Decision Diagram. On the other hand, Jeffrey et al. [40] suggest using a SAT solver for redundancy analysis. Their problem formulation has reduced complexity, allowing for greater performance. However, both strategies cannot detect or resolve policy conflicts. Instead, our approach can also manage this anomaly type, which nonetheless represents the most dangerous one for a firewall configuration.

Cuppens et al. [36] propose an anomaly resolution approach for solving shadowing and redundancy anomalies among intra-firewall policies. A main feature of this study is the definition of a policy rewriting process, which not only solves the aforementioned anomaly types, but can also rewrite a policy in its positive or negative form. The positive form of a policy contains only rules with ‘allow’ actions, whereas the negative form only rules with ‘allow’ actions. Then, the authors extend their model to support also inter-firewall policy analysis [41]. Anyhow, their approach only considers some anomaly classes, while our approach addresses all the possible ones. Besides, their resolution strategy is fully automated, and therefore it also takes decisions that instead should be explicitly addressed by security administrators.

Hu et al. [37], [38] propose a rule-based segmentation technique and a grid-based representation to identify policy anomalies and a policy reordering algorithm for anomaly resolution. They also present a proof-of-concept implementation of a visualization-based firewall policy analysis tool called Firewall Anomaly Management Environment. This study has two main limitations to respect to our proposal. First, the reordering algorithm does not guarantee that all policy conflicts and sub-optimizations are actually removed, because there may be cases where no ordering can be found to achieve a complete anomaly resolution. Second, also this resolution strategy is fully automated, and it may not be always adequate.

In summary, to the best of our knowledge, our approach is characterized by manifold features which overcome existing limitations of the state of the art. First, it proposes a joint
algorithm for firewall policy anomaly analysis and resolution, instead of focusing on a single operation. Second, it aims to solve all possible anomaly types, not only sub-optimizations. Third, it involves human intervention to make decisions that may lead to configuration errors if taken autonomously. Fourth, it reduces the number of queries that the administrator must answer to complete the anomaly resolution.

III. OVERVIEW OF THE APPROACH

In this study, we propose a novel methodology for assisting network administrators in performing intra-firewall policy anomaly detection and resolution. The methodology assists administrators in the latter process by guiding them through the shortest series of well-posed questions to ensure that their original security intents are correctly implemented by the firewall. FIGURE 1 illustrates the logical flow of operations in the proposed approach.

A. FIREWALL POLICY ANOMALY DETECTION

At the beginning of the proposed approach, each pair of firewall rules is analyzed so as to detect if they are afflicted by an anomaly. In fact, anomalies are defined as relations among pairs of firewall rules which may reveal either misconfigurations (i.e., mismatches between administrator intents and the actual policy implemented by the firewall based on its current configuration), or sub-optimizations (i.e., non-minimal firewall configurations, typically due to the presence of redundant rules which do not influence the firewall filtering behavior). The definition of anomalies between pairs of rules follows the traditional definitions and allows for fine-grained anomaly detection while also keeping the computational complexity of the detection algorithm within reasonable levels.

The anomalies that are thus detected are then grouped into two classes. On the one hand, conflict anomalies identify pairs of rules that apply to common packet subsets (i.e., dependent rules) but are associated with conflicting actions. On the other hand, sub-optimization anomalies concern pairs of dependent rules which redundantly process packets in the same way.

The formal definition of the anomaly classes and the formulas used for their identification are detailed in Section V.

B. FIREWALL POLICY ANOMALY RESOLUTION

The task of firewall policy anomaly resolution is composed of four steps: i) firewall rule clustering; ii) sub-optimization resolution; iii) conflict resolution; iv) rule reordering.

1) Firewall rule clustering

After all anomalies have been identified and classified, the firewall rules are grouped in clusters. Two rules are included in the same cluster if they do not have disjoint conditions (i.e., if their conditions match a common subset of packets). Since rules belonging to different clusters are characterized by disjoint conditions, i.e., are matched by different types of traffic, sub-optimizations and conflicts may only arise among pairs of rules in the same cluster. This clustering operation will ease the algorithm defined for firewall policy anomaly resolution, because it can be parallelized and applied to each cluster separately. Grouping rules as clusters is also important, because only considering relations among pairs of firewall rules in most cases does not provide administrators with a clear understanding of the current firewall behavior. Indeed, it may happen that some conflicts regard common regions in the packet space, which are in common to multiple rules and not to a single pair of them.

The formalization of the firewall rule clustering operation is presented in Section VI-A.

2) Sub-optimization resolution

Sub-optimization resolution consists in the automatic resolution of all the sub-optimizations inside each rule cluster. This class of anomalies can be automatically resolved because sub-optimization removal does not change the filtering function implemented by the firewall but only makes its configuration more compact, positively affecting its maintenance and performance.

The algorithm used for the resolution of the sub-optimizations is formalized in Section VI-B.

3) Conflict resolution

Conflict resolution consists in a human-assisted resolution of the conflicts. This task is not fully automated, because
approaches that automatically resolve anomalies following predefined resolution strategies do not allow fine-grained control of the resulting configurations, increasing the risk of ignoring existing errors or even introducing additional ones. As our approach designs also this task in a cluster-based fashion, only clusters containing at least one pair of conflicting rules are examined by the resolution process.

In particular, the approach envisions a sequence of well-posed queries to the administrators with the aim of correcting actual misconfigurations. Two types of queries can be formulated to the administrator: queries of type A and queries of type B.

On the one hand, queries of type A ask the administrator if all the packets matching the condition of a specific rule should be subject to the action of that same rule. If the administrator answers positively to such kind of question, all the conflicts where that rule is involved can be removed, because that is the “winning rule” that must have higher priority than the other involved rules. In this case, some restrictions about the rule relative order are derived, and they will be useful to reorder the rules remaining in the cluster. Besides, under certain conditions, the rules which were deemed not to be the winning ones for a conflict may also be removed.

On the other hand, queries of type B ask the administrator to decide which action between the actions of two rules should be applied to the packets that match both their condition. Queries of type B are the ones that are also used in some techniques for manual conflict resolution. However, a resolution process consisting only or mostly of queries of type B would be unoptimized, because a query would be asked for each specific conflict. Therefore, our approach reduces the number of human interventions, by following a strategy that allows prioritizing queries of type A, whose answers lead to the simultaneous resolution of a higher number of conflicts. The selection of the rules subject to those queries is based on a cost function, related to the number of conflicts that each rule has with the other rules in the previously computed clusters.

In summary, the answers provided by the administrator may be useful both to remove some rules that are not needed anymore after the resolution of the related conflicts, and to define some restrictions about the rule relative order.

The algorithm used for the interaction with the administrator in the context of conflict resolution is formalized in Section VI-C1.

4) Rule reordering

After each cluster has been depured of rules that were deemed useless after the resolution of their conflicts, the remaining rules must be reordered, so as to satisfy the rule order restrictions deriving from the answers themselves.

In our approach, the rule reordering problem has been formulated as a Satisfiability Modulo Theories (SMT) problem. The SMT problem is an enhanced version of the most traditional satisfiability checking technique, the Satisfiability (SAT) problem. A SAT problem is used to check the satisfiability of formulas where only Boolean variables appear. However, for the rule ordering representation, we also need variables and functions of other types (e.g., integer variables to represent the ordering positions). From this point of view, an SMT problem allows checking the satisfiability of formulas in a decidable combination of first-order theories, including arithmetic and uninterpreted functions theories.

An SMT problem can be automatically solved by state-of-the-art efficient solvers, already available in literature [42]–[44]. The solver assigns a firewall rule index to each integer variable, representing a position in the final anomaly-free cluster, and from these assignments the output ordering is derived. The solution of the solver is also guaranteed to be “correct by construction”, as long as the formal models representing the problem components (e.g., firewall rules and their relationship, in our case) capture all the information that may impact the solution correctness. In this way, a human user also has a formal assurance about the result of the automatic sub-processes composing the proposed approach.

The formalization of the SMT problem used to reorder the rules inside each cluster is presented in VI-C2.

Finally, after the rules inside each cluster have been reordered, the algorithm for sub-optimization resolution should be reapplied to ensure that no new sub-optimization was created within the reordering operation. Then, the clusters should be combined to form a single firewall policy. In principle, as there were no anomalies between two rules belonging to different clusters, the relative position of the clusters may be arbitrarily decided, as it is indifferent to the solution correctness. However, the human administrator also has the faculty to order the clusters in a strategic way, e.g., associating higher priority to the cluster whose rules match network packets more frequently, thus optimizing the filtering process of the firewall.

IV. FIREWALL POLICY MODEL

This section recalls the basics of firewall operation and provides a model, derived from the well-established literature on the subject, for describing their configuration. Relations among pairs of firewall rules are also defined and classified.

A. FIREWALLS AND PACKETS

Firewalls filter traversing packets based on the values assumed by a subset of the header fields of the latters (named firewall filtering fields). A field is modeled as a range, i.e., a set of consecutive non-negative integers, $P_n = [a_n, b_n]$, including all values possibly assumed by the field. As an example, IP-layer filtering is typically performed based on the values of 5 IP header fields: source and destination IP address, source and destination port number and transport protocol, which, for IPv4, are modeled, respectively, as ranges $P_{1,2} = [0, 2^{32} - 1]$, $P_{3,4} = [0, 2^{16} - 1]$, and $P_{5} = [0, 2^{8} - 1]$.

For a firewall defined over filtering fields $P_1, P_2, \ldots, P_n$, packets are $N$-tuples $p = (p_1, p_2, \ldots, p_N)$, where $p_n \in P_n$.

In the following, capital letters and the traditional interval notation are used to indicate ranges, e.g., $I = [a, b]$ means that $a, b \in \mathbb{N}$ are respectively the minimum and maximum-valued non-negative integers in range $I$. 

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$n \in [1, N]$, represent the values assumed by the firewall filtering fields in the packet header. If we consider filtering fields to be represented on $N$ orthogonal axes, the set of all possible packets, $P = \mathcal{P} \times \mathcal{P} \times \ldots \times \mathcal{P}$, is a $N$-orthotope (i.e., the generalization of a rectangle in $N$ dimensions) in $N^N$, while each packet is a single point in $N^N$.

A firewall defined over filtering fields $P_1, P_2, \ldots, P_N$ is then a list of rules $fw = \{r_1, r_2, \ldots, r_{|fw|}\}$, where the number of rules, $|fw|$, is the firewall cardinality. A rule $r_i, i \in [1, |fw|]$, is defined as $r_i = (c^i, a^i)$, where $c^i$ is a condition and $a^i$ is an action. For a firewall defined over filtering fields $P_1, P_2, \ldots, P_N$ a condition is a $N$-tuple of ranges $c^i = (C^i_1, C^i_2, \ldots, C^i_N)$, where $C^i_n \subseteq P_n, n \in [1, N]$. A condition implicitly defines a logical predicate over packets $p \in \mathcal{P}$, i.e., $\phi_c(p) = p_1 \in C^i_1 \land p_2 \in C^i_2 \land \ldots \land p_N \in C^i_N$ and identifies the set of packets, $S_c = C^i_1 \times C^i_2 \times \ldots \times C^i_N \subseteq \mathcal{P}$ (also a $N$-orthotope in $N^N$), satisfying the condition itself, i.e., for which predicate $\phi_c(p)$ holds true. Moreover, we assume possible firewall actions to be such that $a^i \in \{allow, deny\}$.

Any time a packet $p = (p_1, p_2, \ldots, p_N)$ reaches a firewall $fw = \{r_1, r_2, \ldots, r_{|fw|}\}$ defined as above, the packet is checked against the firewall rules according to their position in the list. As soon as the first rule whose condition is satisfied by the packet (first matched rule) is found, the rule action is applied and the packet is either forwarded (if the action is allow) or discarded (if, conversely, is deny). The priority of a rule is thus given by the rule position within the firewall list as, whenever a packet satisfies the conditions of multiple rules within a firewall, the applied action is the one of the rule firstly appearing in the list. We assume the priority of a rule $r_i, i \in [1, |fw|]$, within a firewall $fw = \{r_1, r_2, \ldots, r_{|fw|}\}$, to be returned by function $\pi(r_i) = i$, where the lower $\pi(r_i)$ the higher the rule priority.

According to the above description, a firewall can be seen as a function (firewall filtering function) associating to any packet $p \in \mathcal{P}$ one among the possible firewall actions. To make sure that firewall filtering functions are defined over the whole packet set $\mathcal{P}$, the last rule in any firewall (default rule), is supposed to be in the form $r_{|fw|} = (P_1, P_2, \ldots, P_N)\{allow\}$, i.e., to be matched by any packet (completeness hypothesis).

In the following, the traditional dotted decimal notation, with wildcards *, is also used to define ranges referring to IP addresses (e.g., 10.10.10.* stands for 10.10.10.0/8). Moreover, names will be used for protocols (e.g. TCP is a possible protocol type name).

**B. RELATIONS AMONG FIREWALL RULE CONDITIONS**

Here, we explore all possible relations among pairs of rule conditions which, in turn, can straightforwardly be derived from possible relations among pairs of condition components (i.e., ranges).

Given any two ranges $C_n = [a_n, b_n]$ and $C_m = [a_m, b_m]$, they can be seen as two sets of values and, thus, they are in one of the following, mutually exclusive, relations:

- **equivalence (\(=\))**: two ranges are equivalent iff they are the same set, i.e.,
  \[
  C_n = C_m \iff C_n = C_m \iff a_n = a_m \land b_n = b_m \tag{1}
  \]
- **disjointness (\(\perp\))**: two ranges are disjoint iff they are disjoint sets, i.e.,
  \[
  C_n \perp C_m \iff C_n \perp C_m \iff a_n < a_m \lor a_n > b_m \tag{2}
  \]
- **dominance (\(<, >\))**: one range dominates (or is dominated by) the other iff it is a proper superset (or subset) of the other, i.e.,
  \[
  C_n > C_m \iff C_n < C_m \iff (a_n < a_m \land b_n \geq b_m) \lor (a_n = a_m \land b_n > b_m) \tag{3}
  \]
  \[
  C_n < C_m \iff C_n > C_m \iff (a_n > a_m \land b_n \leq b_m) \lor (a_n = a_m \land b_n < b_m) \tag{4}
  \]
- **correlation (\(\sim\))**: two ranges are correlated iff their intersection and their mutual differences are all non-empty, i.e., the ranges are not equivalent, nor disjoint, and neither of them dominates the other:
  \[
  C_n \sim C_m \iff C_n \sim C_m \iff (a_n > a_m \land a_n \leq b_n \land b_n \geq b_m) \lor (b_n < b_m \land b_n \geq a_m \land a_m \leq a_n) \tag{5}
  \]

Similarly to ranges, given any two rule conditions $c^i = (C^i_1, C^i_2, \ldots, C^i_N)$ and $c^j = (C^j_1, C^j_2, \ldots, C^j_N)$, they can be univocally identified with the corresponding packet sets $S_c$ and $S_{c'}$, and are, thus, in one of the following, mutually exclusive, relations:

- **equivalence (\(=\))**: two conditions are equivalent iff the corresponding packet sets are the same set, i.e.,
  \[
  c^i = c^j \iff c^i = c^j \iff C^i_n = C^j_n \forall n \in [1, N] \tag{6}
  \]
- **disjointness (\(\perp\))**: two conditions are disjoint iff the corresponding packet sets are disjoint, i.e.,
  \[
  c^i \perp c^j \iff c^i \perp c^j \iff \exists \bar{n} \in [1, N] \mid C^i_n \perp C^j_n \tag{7}
  \]
- **dominance (\(<, >\))**: one condition dominates (or is dominated by) the other iff the corresponding packet set is a proper superset (or subset) of the packet set of the other, i.e.,
  \[
  c^i > c^j \iff c^i > c^j \iff C^i_n \geq C^j_n \forall n \in [1, N] \land \exists \bar{n} \in [1, N] \mid C^i_n > C^j_n \tag{8}
  \]
  \[
  c^i < c^j \iff c^i < c^j \iff C^i_n \leq C^j_n \forall n \in [1, N] \land \exists \bar{n} \in [1, N] \mid C^i_n < C^j_n \tag{9}
  \]
- **correlation (\(\sim\))**: two conditions are correlated iff their intersection and the mutual differences between the corresponding packet sets are all non-empty, i.e., the conditions are not equivalent, nor disjoint, and neither of them dominates the other:
  \[
  c^i \sim c^j \iff c^i \sim c^j \iff C^i_n \sim C^j_n \forall n \in [1, N] \land \exists \bar{n} \in [1, N] \mid C^i_n \sim C^j_n \tag{10}
  \]
V. FIREWALL POLICY ANOMALY ANALYSIS

As mentioned in the previous sections, this work focuses on firewall anomalies afflicting pairs of rules within the same firewall list (intra-firewall policy anomalies).

Firewall configuration anomalies only arise among pairs of rules \( r_i = (c_i, a_i) \) and \( r_j = (c_j, a_j) \) with non-disjoint conditions \( (c_i \not\subseteq c_j) \), i.e., which are matched by a common subset of packets.

As shown in FIGURE 2, three are two classes of anomalies: conflict anomalies, which involve pairs of rules specifying different actions \( (a_i \neq a_j) \), and sub-optimization anomalies, which, conversely, involve pairs of rules characterized by the same action \( (a_i = a_j) \). The conflict anomalies are divided into three categories: Contradiction, Shadowing Conflict, and Correlation anomalies. Instead, the sub-optimizations anomalies are divided into three different categories: Duplication, Shadowing Redundancy, and Unnecessary anomalies.

For conflict anomalies, only administrators can clarify the intended fate for packets satisfying both rule conditions. Thus, resolving conflict anomalies requires interacting with the administrator, i.e., it cannot be carried out automatically and possibly leads to modifying the firewall filtering function. Conversely, sub-optimization anomalies highlight the presence of redundant rules, unnecessarily stating the same action for a specific kind of traffic. The resolution of sub-optimization anomalies is meant to increase configuration compactness by removing redundancy (which is beneficial to firewall maintenance and filtering performance) and, thus, does not impact the firewall filtering function and can be automatically conducted.

In the following subsections, we will describe the characteristics of each type of anomaly and the formulas to detect them.

1) Conflict anomalies

- **Contradiction anomaly** \( A_{\text{contCon}}(r_i, r_j) \): A contradiction anomaly exists between two rules \( r_i \) and \( r_j \) iff

\[
\pi(r_i) < \pi(r_j) \land c_i > c_j \land a_i \neq a_j
\]

The rule with the lower priority \( (\pi(r_i) < \pi(r_j)) \) is matched by a proper subset of the packets matching the higher priority one \( (c_i > c_j) \), and the rules have different actions \( (a_i \neq a_j) \). This means that the lower priority rule is never actually applied. Note that the contradiction anomaly is symmetric: \( A_{\text{contCon}}(r_i, r_j) \Leftrightarrow A_{\text{contCon}}(r_j, r_i) \).

The resolution of a contradiction anomaly consists in deleting one of the two involved rules.

The set of all contradiction anomalies \( A_{\text{contCon}} \) in \( fw \) is defined as:

\[
\{ \{ r_i, r_j \}, i, j \in [1, |fw|], |A_{\text{contCon}}(r_i, r_j)\} \}
\]

- **Shadowing conflict anomaly** \( A_{\text{shadCon}}(r_i, r_j) \):

A shadowing conflict anomaly exists between two rules \( r_i \) and \( r_j \) iff

\[
c_i = c_j \land a_i \neq a_j
\]

The two rules are matched by exactly the same set of packets \( (c_i = c_j) \), but they indicate different actions \( (a_i \neq a_j) \). This means that the lower priority rule is never actually applied. Note that the shadowing anomaly is symmetric: \( A_{\text{shadCon}}(r_i, r_j) \Leftrightarrow A_{\text{shadCon}}(r_j, r_i) \).

The resolution of a shadowing conflict anomaly consists in deleting one of the two involved rules.

The set of all shadowing conflict anomalies \( A_{\text{shadCon}} \) in \( fw \) is defined as:

\[
\{ \{ r_i, r_j \}, i, j \in [1, |fw|], \{A_{\text{shadCon}}(r_i, r_j)\} \}
\]

- **Correlation anomaly** \( A_{\text{corrCon}}(r_i, r_j) \):

A correlation anomaly exists between two rules \( r_i \) and \( r_j \) iff

\[
c_i \sim c_j \land a_i \neq a_j
\]

The sets of packets matching the two rules have both non-null intersections \( (c_i \sim c_j) \) and differences, and the rules have different actions. Note that the correlation anomaly is symmetric: \( A_{\text{corrCon}}(r_i, r_j) \Leftrightarrow A_{\text{corrCon}}(r_j, r_i) \).

The resolution of a correlation anomaly consists in either the deletion of the lower priority rule or the change of the relative rule order.

The set of all correlation anomalies \( A_{\text{corrCon}} \) in \( fw \) is defined as:

\[
\{ \{ r_i, r_j \}, i, j \in [1, |fw|], \{A_{\text{corrCon}}(r_i, r_j)\} \}
\]
2) Sub-optimization anomalies

- **Duplication anomaly** $A_{\text{duplSub}}(r_i, r_j)$:
  A duplication anomaly exists between two rules $r_i$ and $r_j$ iff:
  \[
  c^i = c^j \land a^i = a^j \land \not\exists r_z | \pi(r_z) < \pi(r_i) < \pi(r_j) \land c^z \not\subseteq c^i \land a^z = a^i \quad (14)
  \]
  The two rules are exactly the same ($c^i = c^j \land a^i = a^j$), if there does not exist another rule in the middle ($\not\exists r_z | \pi(r_z) < \pi(r_i) < \pi(r_j)$), with a different action ($a^z \neq a^i$), that matches a common set of packets of the duplicated rules ($c^z \not\subseteq c^i$). Note that the duplication anomaly is symmetric: $A_{\text{duplSub}}(r_i, r_j) \iff A_{\text{duplSub}}(r_j, r_i)$.

- **Shadowing redundancy anomaly** $A_{\text{shadSub}}(r_i, r_j)$:
  A shadowing redundancy anomaly exists between two rules $r_i$ and $r_j$ iff:
  \[
  \pi(r_i) < \pi(r_j) \land c^i < c^j \land a^i = a^j \land (\not\exists r_z | c^z \not\subseteq c^i \land a^z \neq a^i) \quad (15)
  \]
  The rule with lower priority matches a subset of the packet set matched by the higher priority rule ($\pi(r_i) < \pi(r_j)$), and there does not exist another rule with a different action ($a^z \neq a^i$) that matches a common set of packets of the shadowed rule ($c^z \not\subseteq c^i$). Under these conditions, also this rule relationship is considered a sub-optimization (i.e., it can be dealt with without information loss).

Shadowing redundancy anomalies are automatically resolved by deleting the lower priority rule, which is never actually applied.

The set of all duplication anomalies $A_{\text{duplSub}}$ in $fw$ is defined as $\{\{r_i, r_j\}, i, j \in [1, |fw|], |A_{\text{duplSub}}(r_i, r_j)|\}$.

- **Unnecessary anomaly** ($A_{\text{unneSub}}(r_i, r_j)$):
  An unnecessary anomaly exists between two rules $r_i$ and $r_j$ iff:
  \[
  \pi(r_i) < \pi(r_j) \land c^i \subset c^j \land a^i = a^j \land (\not\exists r_z | c^z \not\subseteq c^i \land a^z \neq a^i) \quad (16)
  \]
  The rule with higher priority matches a subset of the packet set matched by the higher priority rule ($\pi(r_i) < \pi(r_j)$), and it does not exist another rule with a different action ($a^z \neq a^i$) that matches a common set of packets of the smaller rule ($c^z \not\subseteq c^i$). Under these conditions, also this rule relationship is considered a sub-optimization (i.e., it can be dealt with without information loss).

Unnecessary anomalies are automatically resolved by deleting the higher priority rule, whose presence does not impact the firewall filtering function.

The set of all unnecessary anomalies $A_{\text{unneSub}}$ in $fw$ is defined as $\{\{r_i, r_j\}, i, j \in [1, |fw|], |A_{\text{unneSub}}(r_i, r_j)|\} \cup A_{\text{unneSub}(r_j, r_i)}$.

### VI. FIREWALL ANOMALY RESOLUTION

In this section, all the steps of the proposed methodology for firewall anomaly resolution are described. Specifically, the aim of the approach is to assist network administrators in resolving policy anomalies. In doing so, it executes some steps automatically, while for others, it requires interaction with the administrator, i.e., their answer to a series of well-posed queries.

In accordance with the diagram in Figure 1, the input of the resolution algorithm is a firewall configuration $fw = (r_1, r_2, \ldots, r_{|fw|})$, defined as in Section IV, and the sets of all configuration anomalies IV, i.e., $A_{\text{contCon}}, A_{\text{shadCon}}, A_{\text{corrCon}}, A_{\text{duplSub}}, A_{\text{shadSub}}$ and $A_{\text{unneSub}}$, already identified with the strategy discussed in Section V.

### A. RULE CLUSTER COMPUTATION

As a first step, a partition $G = \{g_1, g_2, \ldots, g_L\} \subseteq 2^{r_1, r_2, \ldots, r_{|fw|}}$ of the set of the firewall rules is defined.

Specifically, elements $g_\ell \in G, \ell \in [1, L]$, are all and only clusters of rules within $fw$, i.e., non-empty subsets of the set of firewall rules, which satisfy both the following properties:

\[
\begin{align*}
(i) & \quad \forall i, j \in [1, |fw|], i \neq j, g_\ell \subseteq G, r_\ell = \{c^i, a^i\}, \\
& \quad r_j = \{c^j, a^j\} \in g_\ell \Rightarrow c^j \not\subseteq c^i \\
(ii) & \quad \forall i, j \in [1, |fw|], i \neq j, g_\ell, g_\ell \subseteq G, r_\ell = \{c^i, a^i\} \subseteq g_\ell, \\
& \quad r_j = \{c^j, a^j\} \in g_\ell \Rightarrow c^j \not\subseteq c^i
\end{align*}
\]

Properties (17) state that clusters indicate all and only groups of rules belonging to $fw$ with non-disjoint conditions. Clusters $g_\ell \subseteq G, \ell \in [1, L]$ can be separately examined with respect to the detection and resolution of configuration anomalies. Indeed, since rules belonging to different clusters are characterized by disjoint conditions, i.e., are matched by different types of traffic, conflicts may only arise among pairs of rules in the same cluster. Moreover, the resolution of a conflict involving rules in a cluster does not affect rules in the others. For these reasons, the second step of our methodology consists in computing, for each $g_\ell \subseteq G, \ell \in [1, L]$, the subsets $A_{\text{contCon}, \ell}, A_{\text{shadCon}, \ell}, A_{\text{corrCon}, \ell}, A_{\text{duplSub}, \ell}, A_{\text{shadSub}, \ell}$ and $A_{\text{unneSub}, \ell}$ grouping anomalies of each type only involving rules in cluster $g_\ell$.

The advantages of creating this partition are manifold.

On the one hand, human administrators can more intuitively understand how each rule is related with the other ones, and therefore they may have a clearer understanding of the firewall behavior. On the other hand, as there are no anomalies between two rules belonging to different clusters, the remainder of the resolution algorithm can be parallelized, thus improving the overall efficiency of the process.

Once this step is concluded, the resolution of anomalies in each cluster is separately conducted with a positive impact.
Algorithm 1 for sub-optimization resolution

Input: the cluster $g_{\ell}$, the sub-optimization sets $A_{\text{duplSub},\ell}$, $A_{\text{shadSub},\ell}$, the conflict sets $A_{\text{contCon},\ell}$, $A_{\text{shadCon},\ell}$, $A_{\text{corrCon},\ell}$

Output: the modified cluster $g_{\ell}$, the modified sets $A_{\text{contCon},\ell}$, $A_{\text{shadCon},\ell}$, $A_{\text{corrCon},\ell}$

1: $R_{\text{toRemove}} \leftarrow \emptyset$
2: for each $(r_1, r_2) \in (A_{\text{duplSub},\ell} \cup A_{\text{shadSub},\ell})$ do
3:   if $\pi(r_1) > \pi(r_2)$ then
4:     $R_{\text{toRemove}} \leftarrow R_{\text{toRemove}} \cup \{r_1\}$
5:     $g_{\ell} \leftarrow g_{\ell} \setminus \{r_1\}$
6:   else
7:     $R_{\text{toRemove}} \leftarrow R_{\text{toRemove}} \cup \{r_2\}$
8:     $g_{\ell} \leftarrow g_{\ell} \setminus \{r_2\}$
9: for each $(r_1, r_2) \in A_{\text{unneSub},\ell}$ do
10:   if $\pi(r_1) < \pi(r_2)$ then
11:     $R_{\text{toRemove}} \leftarrow R_{\text{toRemove}} \cup \{r_1\}$
12:     $g_{\ell} \leftarrow g_{\ell} \setminus \{r_1\}$
13:   else
14:     $R_{\text{toRemove}} \leftarrow R_{\text{toRemove}} \cup \{r_2\}$
15:     $g_{\ell} \leftarrow g_{\ell} \setminus \{r_2\}$
16: for each $r_1 \in R_{\text{toRemove}}$ do
17:   for each $A_{X,\ell} \in \{A_{\text{contCon},\ell}, A_{\text{shadCon},\ell}, A_{\text{corrCon},\ell}\}$ do
18:     for each $(r_m, r_n) \in A_{X,\ell}$ do
19:       if $r_k = r_m \lor r_k = r_n$ then
20:         $A_{X,\ell} \leftarrow A_{X,\ell} \setminus \{r_m, r_n\}$
21: return $g_{\ell}$, $A_{\text{contCon},\ell}$, $A_{\text{shadCon},\ell}$, $A_{\text{corrCon},\ell}$


1) Queries for conflict resolution
Our approach envisions two types of queries that can be posed to the network administrators:

1) Queries of “type A” are in the form:

$$queryTypeA = (r_i, r_j)$$

where the semantics of the query is as follows

Shall set of packets $\{p \in P \mid p \text{ satisfies } c^i\}$ be subject to action $a^i$?

In other words, the administrator is asked if the action $a^i$ of rule $r_i$ should actually be applied to all the packets matching the condition $c^i$ of $r_i$. The possible answers the administrator can provide to the query are “yes” or “no”. If the administrator answers “yes”, all the conflicts where $r_i$ is involved can be solved, as it has been established that $a^i$ must be applied to all packets matching $c^i$. In particular, in the final anomaly-free policy, $r_i$ must have higher priority than the rules it had conflicts with. Those rules may also be removed, if they have become useless after conflict resolution. Instead, if the administrator answers “no”, then some packets matching $c^i$ must not be subject to the action $a_i$. This means that there is another rule, in an anomalous relationship with $r_i$, that must have higher priority than $r_i$.

2) Queries of “type B” are in the form:

$$queryTypeB = (r_i, r_j)$$

where the semantics of the query is as follows

Shall set of packets $\{p \in P \mid p \text{ satisfies both } c^i \text{ and } c^j\}$ be subject to action $a^i$ or $a^j$?

In other words, the administrator is asked which action, between $a^i$ and $a^j$, should be applied to the packet space which originated a conflict among $r_i$ and $r_j$. The possible answers the administrator can provide to the query are $a^i$ or $a^j$. If the administrator answers $a^i$, then in the final anomaly-free policy $r_i$ must have higher priority than $r_j$ (or vice-versa, if the administrator answers $a^j$).

Queries of type A may lead to solving multiple conflicts with a single answer, thus reducing the total number of queries the administrator must answer and optimizing the process as the human operator is assisted in the conflict resolution. Queries of type B are asked only when it is not possible to ask queries of type A anymore. This happens when all queries of type A whose answers may lead to a simultaneous resolution of multiple conflicts have already been asked (independently of the answers), and all remaining conflicts only involve pairs of rules. However, it is expected that a large number of conflicts are already solved before the process starts to ask queries of type B to the administrator. This feature represents an optimization with respect to traditional approaches for firewall policy conflict resolution, where only variants of

B. SUB-OPTIMIZATION ANOMALY RESOLUTION
The second step of the proposed resolution methodology consists in the automatic resolution of all detected sub-optimization anomalies according to the strategy formalized in Algorithm 1. In words, this algorithm fulfills the following resolution policies:

- $\forall (r_i, r_j) \in (A_{\text{duplSub},\ell} \cup A_{\text{shadSub},\ell})$:
  1) delete the lower priority rule in the pair from the firewall configuration;

- $\forall (r_i, r_j) \in A_{\text{unneSub},\ell}$:
  1) delete the higher priority rule in the pair from the firewall configuration;

- for any deleted rule, delete all conflict anomalies in $(A_{\text{contCon},\ell} \cup A_{\text{shadCon},\ell} \cup A_{\text{corrCon},\ell})$ involving the rule.

C. CONFLICT ANOMALY RESOLUTION
Once all firewall sub-optimization anomalies have been resolved and conflict anomalies have been updated accordingly, the resolution of firewall conflict anomalies needs to take place. This task is composed of two main operations: 1) a series of well-posed queries to the network administrator, with the objective of establishing constraints about rule priority or rule deletion for the resolution of all anomalies; 2) the resolution of an SMT problem to establish the actual priority of the remaining rule, so that the output rule set is conflict-free.
Algorithm 2 for conflict resolution

Input: the cluster \( g_e \), the conflict sets \( A_{contCon, t} \), \( A_{shadCon, t} \), \( A_{corrCon, t} \)

Output: the modified cluster \( g_e \), the set of constraints \( K \)

1: for each \( r_e \) in \( g_e \) do
2: toAsk[\( r_e \)] ← true
3: endFirstPart ← false
4: while endFirstPart = false do
5: conflictNo ← compConfNo(\( g_e \), \( A_{contCon, t} \), \( A_{shadCon, t} \), \( A_{corrCon, t} \))
6: \( g_t \leftarrow orderByDescConflictNo(g_e, conflictNo) \)
7: for each \( r_i \) in \( g_t \) do
8: if conflictNo[\( r_i \)] ≤ 1 then
9: endFirstPart ← true
10: break
11: if toAsk[\( r_i \)] = false then
12: continue
13: response ← queryTypeA(\( r_i \))
14: toAsk[\( r_i \)] ← false
15: if response = “yes” then
16: for each \( \{ r_j, r_i \} \in A_{contCon, t} \lor \{ r_i, r_j \} \in A_{shadCon, t} \) do
17: if \( r_j = r_m \lor r_j = r_n \) then
18: \( A_{X, t} \leftarrow \{ A_{X, t} \} \)
19: for each \( \{ r_j, r_i \} \in A_{corrCon, t} \lor \{ r_i, r_j \} \in A_{shadCon, t} \) do
20: toAsk[\( r_i \)] ← false
21: \( A_{corrCon, t} \leftarrow A_{corrCon, t} \setminus \{ r_i, r_j \} \)
22: \( A_{shadCon, t} \leftarrow A_{shadCon, t} \setminus \{ r_i, r_j \} \)
23: \( k_0 \leftarrow \{ greater(i, j) = true \} \)
24: \( K \leftarrow K \cup \{ k_0 \} \)
25: for each \( \{ r_i, r_j \} \in A_{corrCon, t} \lor A_{shadCon, t} \lor A_{corrCon, t} \) do
26: response ← queryTypeB(\( r_i, r_j \))
27: if response = \( a' \) then
28: \( k_0 \leftarrow greater(i, j) = true \)
29: else
30: \( k_0 \leftarrow greater(j, i) = true \)
31: \( K \leftarrow K \cup \{ k_0 \} \)
32: return \( g_t \), \( K \)

queries of type B are asked to the network administrator, who therefore must solve the conflicts one by one.

The algorithm works on the rule of a single cluster \( g_t \), after sub-optimizations have been removed as previously described. Therefore, the algorithm can be executed in parallel on each cluster, thus improving the efficiency of the conflict resolution process. The conflict resolution methodology proposed in this paper is detailed in Algorithm 2.

In the beginning, the algorithm initializes some variables and associative arrays (lines 1-3). In greater detail, for each rule \( r_k \) of the analyzed cluster, the corresponding value of the associative array toAsk[\( r_k \)] is set to true: this means that conflicts related to that rule still require queries to be posed to the administrator. Instead, the boolean variable endFirstPart is set to false, and it will be used in termination conditions.

After this preliminary phase, the core of the algorithm is composed of two main blocks: in the former (lines 4-34) the administrator is asked queries of type A, in the latter (lines 35-41) he is asked queries of type B to solve the remaining conflicts.

The first block (lines 4-34) has an iterative structure, as it is repeatedly executed until endFirstPart is not modified to true. In each iteration, the number of conflicts that are afflicting the rules that are still present in the cluster \( g_t \) is computed with the compConfNo function, which is detailed in Algorithm 3. This function simply checks if any rule \( r_e \) appears in the rule pairs composing the three sets \( A_{contCon, t} \), \( A_{shadCon, t} \), \( A_{corrCon, t} \), and sets the value of the associative array conflictNo[\( r_e \)] to the total number of current conflicts for that rule. Then the algorithm reorders the rules in \( g_t \) in descending order of conflict number. The quick sort algorithm, well-known in literature, is employed for this reordering operation. At that point, a single rule \( r_i \) of the reordered cluster is analyzed at a time. If this rule has only one conflict or it does not have any conflict (i.e., conflictNo[\( r_i \)] ≤ 1), the first part of the algorithm must conclude (i.e., the boolean variable endFirstPart is set to true), because it is not possible anymore to solve multiple conflicts at the same time asking single queries of type A. Otherwise, if asking a question for that rule is still useful (i.e., toAsk[\( r_i \)] = true), a query of type A is posed to the administrator, asking if the action \( a' \) must be applied to all packets matching the condition \( c' \). Independently of the answer, toAsk[\( r_i \)] is set to false, because the query has been asked for that rule. Then, a positive answer means that the rule \( r_i \) wins all the conflicts where it is involved, and those conflicts can be automatically removed from their respective sets \( A_{contCon, t} \), \( A_{shadCon, t} \), \( A_{corrCon, t} \). Their simultaneous removal with a single query represents a strong point of the proposed methodology, as it avoids specific queries for each conflict. In addition to removing them from those sets, other operations must be performed to ensure the conflict removal in the final firewall policy that will be produced at the end of the whole resolution algorithm. On the one hand, if there was a contradiction between \( r_i \) and another rule \( r_j \), or if there was a shadowing conflict where \( r_i \) shadowed \( r_j \), then \( r_j \) can be directly deleted from the cluster \( g_t \), and all the conflicts where it was involved removed from the sets \( A_{contCon, t} \), \( A_{shadCon, t} \), \( A_{corrCon, t} \). On the other hand, if there was a correlation between \( r_i \) and another rule \( r_j \), or if there was a shadowing conflict where \( r_i \) was shadowed by \( r_j \), \( r_j \) cannot be directly removed. Instead, a constraint is created for the SMT problem (i.e., \( greater(i, j) = true \)), stating that \( r_j \) must have higher priority than \( r_i \) in the final anomaly-free policy. More details about the role of these constraints will be explained in Section VI-C2. After all these operations, the whole first block is repeated until the previously explained termination condition is verified.

The second block (lines 35-41) is more similar to traditional conflict resolution strategies. For each conflict between
Algorithm 3 for conflict number computation

**Input:** the cluster \( g_\ell \), the conflict sets \( \mathcal{A}_{\text{corrCon,} \ell} \), \( \mathcal{A}_{\text{shadCon,} \ell} \), \( \mathcal{A}_{\text{contCon,} \ell} \)  

**Output:** the associative array conflictNo, associating the conflict number for each rule in \( g_\ell \)

1. for each \( r_k \) in \( g_\ell \) do  
2. \( \text{conflictNo}[r_k] \leftarrow 0 \)  
3. for each \( (r_i, r_j) \in \mathcal{A}_{\text{contCon,} \ell} \cup \mathcal{A}_{\text{shadCon,} \ell} \cup \mathcal{A}_{\text{corrCon,} \ell} \) do  
4. \( \text{conflictNo}[r_i] \leftarrow \text{conflictNo}[r_i] + 1 \)  
5. \( \text{conflictNo}[r_j] \leftarrow \text{conflictNo}[r_j] + 1 \)  
6. return conflictNo

Two classes of standard constraints that are always included in the SMT problem, independently of the decisions taken by the administrator in Algorithm 2, are the following:

1. The integer variables \( x_1, x_2, \ldots, x_{|g_\ell|} \) can only be assigned with values corresponding to the subscripts of the rules included in \( |g_\ell| \). This requirement is enforced by introducing the following constraint class in the SMT problem, limiting the solution space to those specific integer values:

   \[
   \forall i \in \{1, 2, \ldots, |g_\ell|\} \; \exists j \in \ell \in g_\ell \; (x_i = j)
   \]

2. For each pair of variable \( x_i \) and \( x_j \), we assume that the rule whose index is assigned to \( x_i \) has higher priority than the rule whose index is assigned to \( x_j \) in the final reordered cluster \( g'_\ell \). This assumption is guaranteed by introducing the following constraint class in the SMT problem, generalized for any pair of variables:

   \[
   \forall i, j \; i = \{1, 2, \ldots, |g_\ell|\}, j = \{1, 2, \ldots, |g_\ell|\}, i < j, \; \text{greater}(x_i, x_j) = \text{true}
   \]

In addition to these basic classes, all the constraints of the set \( K \) computed with Algorithm 2 are included in the formulated SMT problem. Those constraints reflect priority relationships between rules as a consequence of the conflict resolution decisions taken by the network administrator. Besides, for any pair of rules \( r_i \) and \( r_j \) belonging to \( g_\ell \) such that no constraint exists in \( K \) (e.g., because no conflict was present between them), then a simple constraint on the greater predicate is included so as to preserve their original mutual ordering, i.e., \( \text{greater}(i, j) = \text{true} \) if \( r_i \) had higher priority, \( \text{greater}(i, j) = \text{false} \) otherwise.

An automated SMT solver is then used to find a correct solution to the problem. If any, the solver assigns a specific integer number among the subscripts of the rules in \( g_\ell \) to each variable, and this information is easily used to reorder all rules in the output cluster \( g'_\ell \).

Finally, as the rule reordering determined by constraints of the SMT problem did not take into account how sub-optimizations were previously removed, Algorithm 1 should simply be applied to the final \( g'_\ell \), to remove any possible sub-optimization caused by the new order. Then, all clusters can be combined either in an arbitrary way or with a strategic ordering. The former is always a feasible choice, because pairs of rules of different clusters do not have any anomaly. The latter may be possible under specific circumstances, e.g., the administrator knows which rules are matched by network packets the most, and decides to assign higher priority to their cluster.
D. TIME COMPLEXITY ANALYSIS

Here we discuss the worst-case time complexity analysis of the algorithm formalized for firewall anomaly resolution. As previously mentioned, this algorithm is composed of two main sub-algorithms: the first one for sub-optimization anomaly resolution (Algorithm 1), the second one for conflict anomaly resolution (Algorithm 2, which internally calls Algorithm 3 as a component). For their time complexity analysis, the interactions with the user have not been considered.

The worst-case time complexity of Algorithm 1, used for sub-optimization resolution in a single cluster, can be estimated as the sum of the time complexities of two main sequential code blocks. Lines 2-15 have \( O(\ell_\text{dulpSub},_t \cup \ell_\text{shadSub},_t \cup \ell_\text{anneSub},_t) \) complexity, because \( O(1) \) operations are performed on each element of the three anomaly sets sequentially. In fact, even if lines 9-8 are the same for the elements of two sets, they are still executed sequentially on them. Lines 16-20 have \( O(|gr| \cdot (\ell_\text{dulpSub},_t \cup \ell_\text{shadSub},_t \cup \ell_\text{anneSub},_t)) \) complexity. The difference with respect to the previous code block is that the second one consists of two nested loops. The external one iterates on each element of \( R_{\text{Remove}} \), which coincides with the \( g_t \) set (except for an element) in the worst case, whereas the internal one requires a number of iterations that is again equal to the number of elements of the three sub-optimization sets. All the other lines of the algorithm are \( O(1) \) operations. Summing up, the overall worst-case time complexity for sub-optimization resolution is equal to the time complexity of the second code block, i.e.,

\[
O(|gr| \cdot (\ell_\text{dulpSub},_t \cup \ell_\text{shadSub},_t \cup \ell_\text{anneSub},_t))
\]

because that term is dominant with respect to the other one, where the \( g_t \) was not involved.

The worst-complexity of Algorithm 2, which also embeds Algorithm 3, used for conflict resolution in a single cluster, can be estimated as the sum of the time complexities of three sequential code blocks. The first block is the less impacting on the overall time complexity. Indeed, lines 1-2 have \( O(|g_t|) \) complexity, because a single \( O(1) \) operation is performed on each element of the modified cluster. Instead, the second block is the most impacting, because it also includes the whole Algorithm 3. This second block, composed of lines 4-34, is a loop of three sub-blocks on all the elements of \( g_t \) in the worst case:

1) line 6 is a call to Algorithm 3, whose worst-case time complexity is \( O(|g_t|) + O(\ell_\text{contCon},_t \cup \ell_\text{shadCon},_t \cup \ell_\text{corrCon},_t) \) because it is composed of two sequential loops one on the rules of the cluster, the other one on the elements of the three conflict sets;
2) line 7 is a call to the quick sort algorithm for rule reordering, and therefore it has \( O(|g_t| \cdot \log(|g_t|)) \);
3) lines 8-34 is another nested loop, and has a \( O(|g_t| \cdot (\ell_\text{contCon},_t \cup \ell_\text{shadCon},_t \cup \ell_\text{corrCon},_t)^2) \) complexity in the worst case, that occurs when the condition in line 16 is true.

Therefore, by summing the different contributions and adding the multiplying factor due to the external loop, the time complexity related to the second block has an overall \( O(|g_t|)^2 \cdot (\log(|g_t|) + |\ell_\text{contCon},_t \cup \ell_\text{shadCon},_t \cup \ell_\text{corrCon},_t|) \) complexity. Then, the third block, composed of lines 35-41, simply have \( O(|\ell_\text{contCon},_t \cup \ell_\text{shadCon},_t \cup \ell_\text{corrCon},_t|) \) complexity, as it is composed of a single loop on the elements of the three sets. Summing up, the time complexity of the whole algorithm is the one of the second block, because it is the dominant term for the asymptotic complexity.

VII. IMPLEMENTATION AND CASE STUDY EVALUATION

The proposed optimized approach has been implemented as a Java-based framework, which offers an interactive view for human administrators to read the questions that are posed by the methodology, and to answer them so that the approach can subsequently resolve the conflicts. The tool interacts with the Java APIs of Z3, a state-of-the-art theorem prover developed by Microsoft Research [43], for the resolution of the SMT problem.

The framework has been validated on a machine with an Intel i7-6700 CPU at 3.40 GHz and 32GB of RAM. Specifically, it has been evaluated on a series of use cases, so as to assess if the produced result is actually compliant with the objectives of the human administrators. Here, we describe a complete use case, which was useful to stress the defined algorithm in the operations of anomaly resolution, because all anomaly types are included.

TABLE 1 reports a list of rules composing a firewall configuration. This list size has been chosen for multiple reasons. On the one hand, the scope of our proposal consists in intra-firewall anomaly resolution, a scenario where the number of rules is limited because otherwise humans would not be able to manage all of them by themselves. On the other hand, the role of this validating case study is also to explain how the proposed approach works in all the phases. For sake of visualization simplicity, the conditions of all rules specified in TABLE 1 only select packets depending on the values of their source and destination IP addresses. In this way, it will be possible to graphically represent their packet spaces as rectangles in a 2-space, instead of penteracts in a 5-space.

Initially, the anomaly analysis algorithm identifies all anomalies (i.e., sub-optimizations and conflicts) among the rules of TABLE 1, classifying them into the sets \( \ell_\text{contCon}, \), \( \ell_\text{shadCon}, \), \( \ell_\text{corrCon}, \), \( \ell_\text{dulpSub}, \), \( \ell_\text{shadSub}, \) and \( \ell_\text{anneSub}, \) according to the anomaly classification metrics discussed in Section V:

- \( \ell_\text{contCon} = \{ r_6, r_{11} \} \);
- \( \ell_\text{shadCon} = \{ r_4, r_{13} \} \);
- \( \ell_\text{corrCon} = \{ r_1, r_2, \{ r_1, r_{11} \}, \{ r_1, r_{17} \}, \{ r_2, r_{14} \}, \{ r_3, r_4, r_5 \}, \{ r_3, r_9 \}, \{ r_4, r_{14} \}, \{ r_7, r_{16} \}, \{ r_8, r_{16} \} \};
- \( \ell_\text{dulpSub} = \{ r_5, r_9 \} \);
- \( \ell_\text{shadSub} = \{ r_4, r_{15} \} \);
- \( \ell_\text{anneSub} = \{ r_{10}, r_{14} \} \).

After the anomaly analysis is completed, the rules are clustered, so that any two rules belonging to the same cluster have non-disjoint conditions:

- \( g_1 = \{ r_1, r_2, r_3, r_4, r_5, r_6, r_9, r_{10}, r_{11}, r_{13}, r_{14}, r_{15}, r_{17} \} \);
TABLE 1: Initial firewall policy of the use case

<table>
<thead>
<tr>
<th>Rule</th>
<th>Source IP Address</th>
<th>Destination IP Address</th>
<th>Action</th>
<th>Priority</th>
</tr>
</thead>
<tbody>
<tr>
<td>r1</td>
<td>[124.60.0.22, 124.60.0.32]</td>
<td>[152.88.36.14, 124.60.0.78]</td>
<td>deny</td>
<td>7</td>
</tr>
<tr>
<td>r2</td>
<td>[124.60.0.16, 124.60.0.72]</td>
<td>[152.88.36.64, 124.60.0.92]</td>
<td>allow</td>
<td>2</td>
</tr>
<tr>
<td>r3</td>
<td>[124.60.0.134, 124.60.0.146]</td>
<td>[152.88.36.42, 124.60.0.58]</td>
<td>deny</td>
<td>3</td>
</tr>
<tr>
<td>r4</td>
<td>[124.60.0.84, 124.60.0.146]</td>
<td>[152.88.36.32, 124.60.0.100]</td>
<td>allow</td>
<td>4</td>
</tr>
<tr>
<td>r5</td>
<td>[124.60.0.130, 124.60.0.140]</td>
<td>[152.88.36.74, 124.60.0.126]</td>
<td>deny</td>
<td>5</td>
</tr>
<tr>
<td>r6</td>
<td>[124.60.0.26, 124.60.0.62]</td>
<td>[152.88.36.24, 124.60.0.32]</td>
<td>deny</td>
<td>6</td>
</tr>
<tr>
<td>r7</td>
<td>[124.60.0.162, 124.60.0.192]</td>
<td>[152.88.36.124, 124.60.0.132]</td>
<td>allow</td>
<td>7</td>
</tr>
<tr>
<td>r8</td>
<td>[124.60.0.186, 124.60.0.214]</td>
<td>[152.88.36.88, 124.60.0.108]</td>
<td>allow</td>
<td>8</td>
</tr>
<tr>
<td>r9</td>
<td>[124.60.0.130, 124.60.0.140]</td>
<td>[152.88.36.74, 124.60.0.126]</td>
<td>allow</td>
<td>9</td>
</tr>
<tr>
<td>r10</td>
<td>[124.60.0.76, 124.60.0.80]</td>
<td>[152.88.36.120, 124.60.0.130]</td>
<td>deny</td>
<td>10</td>
</tr>
<tr>
<td>r11</td>
<td>[124.60.0.26, 124.60.0.62]</td>
<td>[152.88.36.24, 124.60.0.32]</td>
<td>allow</td>
<td>11</td>
</tr>
<tr>
<td>r12</td>
<td>[124.60.0.180, 124.60.0.196]</td>
<td>[152.88.36.20, 124.60.0.30]</td>
<td>deny</td>
<td>12</td>
</tr>
<tr>
<td>r13</td>
<td>[124.60.0.110, 124.60.0.120]</td>
<td>[152.88.36.28, 124.60.0.36]</td>
<td>deny</td>
<td>13</td>
</tr>
<tr>
<td>r14</td>
<td>[124.60.0.62, 124.60.0.138]</td>
<td>[152.88.36.82, 124.60.0.114]</td>
<td>deny</td>
<td>14</td>
</tr>
<tr>
<td>r15</td>
<td>[124.60.0.94, 124.60.0.102]</td>
<td>[152.88.36.54, 124.60.0.60]</td>
<td>allow</td>
<td>15</td>
</tr>
<tr>
<td>r16</td>
<td>[124.60.0.180, 124.60.0.188]</td>
<td>[152.88.36.92, 124.60.0.126]</td>
<td>deny</td>
<td>16</td>
</tr>
<tr>
<td>r17</td>
<td>[124.60.0.10, 124.60.0.24]</td>
<td>[152.88.36.24, 124.60.0.32]</td>
<td>allow</td>
<td>17</td>
</tr>
</tbody>
</table>

FIGURE 3: Visual representation of the clusters

- $g_2 = \{r_7, r_8, r_{16}\}$
- $g_3 = \{r_{12}\}$

The result of the clustering operation is also graphically shown in FIGURE 3.

After all rules have been clustered, the anomaly resolution algorithm is applied to each cluster independently from the other ones, as it can be parallelized.

The cluster with the higher number of rules and anomalies is $g_1$. For $g_1$, the six sets $\mathcal{A}_{\text{contCon},1}$, $\mathcal{A}_{\text{shadCon},1}$, $\mathcal{A}_{\text{corrCon},1}$, $\mathcal{A}_{\text{duplSub},1}$, $\mathcal{A}_{\text{shadSub},1}$ and $\mathcal{A}_{\text{unneSub},1}$ are derived from the original sets $\mathcal{A}_{\text{contCon}}$, $\mathcal{A}_{\text{shadCon}}$, $\mathcal{A}_{\text{corrCon}}$, $\mathcal{A}_{\text{duplSub}}$, $\mathcal{A}_{\text{shadSub}}$ and $\mathcal{A}_{\text{unneSub}}$, by keeping only the anomalies that involve rules of $g_1$. Specifically:

- $\mathcal{A}_{\text{contCon},1} = \{\{r_6, r_{11}\}\}$
- $\mathcal{A}_{\text{shadCon},1} = \{\{r_4, r_{15}\}\}$
- $\mathcal{A}_{\text{corrCon},1} = \{\{r_1, r_2\}, \{r_1, r_{11}\}, \{r_1, r_{17}\}, \{r_2, r_{14}\}, \{r_3, r_4\}, \{r_4, r_5\}, \{r_4, r_{14}\}\}$
- $\mathcal{A}_{\text{duplSub},1} = \{\{r_5, r_9\}\}$
- $\mathcal{A}_{\text{shadSub},1} = \{\{r_4, r_{15}\}\}$
- $\mathcal{A}_{\text{unneSub},1} = \{\{r_{10}, r_{14}\}\}$

First, the sub-optimization resolution strategy, previously formalized in Algorithm 1, is applied to all the sub-optimizations included in the sets $\mathcal{A}_{\text{duplSub},1}$, $\mathcal{A}_{\text{shadSub},1}$ and $\mathcal{A}_{\text{unneSub},1}$:

- as the sub-optimization between $r_5$ and $r_9$ is a duplication anomaly, one of the two rules is removed, for simplicity the one with lower priority, i.e., $r_9$;
- as the sub-optimization between $r_4$ and $r_{15}$ is a shadowing redundancy, the rule with lower priority is removed, i.e., $r_{15}$;
- as the sub-optimization between $r_{10}$ and $r_{14}$ is an unnecessary anomaly, the rule with higher priority is removed, i.e., $r_{10}$.

After these removal operations, the cluster $g_1$ has become as follows: $g_1 = \{r_1, r_2, r_3, r_4, r_5, r_6, r_{11}, r_{13}, r_{14}, r_{16}\}$. Instead, the three conflict sets have been modified as follows: $\mathcal{A}_{\text{contCon},1} = \{\{r_6, r_{11}\}\}$, $\mathcal{A}_{\text{shadCon},1} = \{\{r_4, r_{13}\}\}$, $\mathcal{A}_{\text{corrCon},1} = \{\{r_1, r_{12}\}, \{r_1, r_{11}\}, \{r_1, r_{17}\}, \{r_2, r_{14}\}, \{r_3, r_4\}, \{r_4, r_5\}, \{r_4, r_{14}\}\}$.

Second, the conflict resolution strategy, previously for-
TABLE 2: Output firewall policy of the use case

<table>
<thead>
<tr>
<th>Rule</th>
<th>Source IP Address</th>
<th>Destination IP Address</th>
<th>Action</th>
<th>Priority</th>
</tr>
</thead>
<tbody>
<tr>
<td>r4</td>
<td>[124.60.0.84, 124.60.0.146]</td>
<td>[152.88.36.22, 124.60.0.100]</td>
<td>allow</td>
<td>1</td>
</tr>
<tr>
<td>r3</td>
<td>[124.60.0.134, 124.60.0.146]</td>
<td>[152.88.36.42, 124.60.0.58]</td>
<td>deny</td>
<td>2</td>
</tr>
<tr>
<td>r5</td>
<td>[124.60.0.130, 124.60.0.140]</td>
<td>[152.88.36.74, 124.60.0.126]</td>
<td>deny</td>
<td>3</td>
</tr>
<tr>
<td>r14</td>
<td>[124.60.0.62, 124.60.0.98]</td>
<td>[152.88.36.82, 124.60.0.144]</td>
<td>deny</td>
<td>4</td>
</tr>
<tr>
<td>r1</td>
<td>[124.60.0.22, 124.60.0.32]</td>
<td>[152.88.36.14, 124.60.0.76]</td>
<td>deny</td>
<td>5</td>
</tr>
<tr>
<td>r2</td>
<td>[124.60.0.16, 124.60.0.72]</td>
<td>[152.88.36.64, 124.60.0.92]</td>
<td>allow</td>
<td>6</td>
</tr>
<tr>
<td>r11</td>
<td>[124.60.0.26, 124.60.0.62]</td>
<td>[152.88.36.24, 124.60.0.32]</td>
<td>allow</td>
<td>7</td>
</tr>
<tr>
<td>r17</td>
<td>[124.60.0.10, 124.60.0.24]</td>
<td>[152.88.36.24, 124.60.0.32]</td>
<td>allow</td>
<td>8</td>
</tr>
<tr>
<td>r16</td>
<td>[124.60.0.180, 124.60.0.188]</td>
<td>[152.88.36.92, 124.60.0.126]</td>
<td>deny</td>
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</tr>
<tr>
<td>r7</td>
<td>[124.60.0.162, 124.60.0.192]</td>
<td>[152.88.36.124, 124.60.0.132]</td>
<td>allow</td>
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</tr>
<tr>
<td>r8</td>
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<td>[152.88.36.88, 124.60.0.108]</td>
<td>allow</td>
<td>11</td>
</tr>
<tr>
<td>r12</td>
<td>[124.60.0.180, 124.60.0.196]</td>
<td>[152.88.36.20, 124.60.0.30]</td>
<td>deny</td>
<td>12</td>
</tr>
</tbody>
</table>

malized in Algorithm 2, is applied to solve all the conflicts throughout an interaction with the human administrator.

In the first iteration of the first code block of Algorithm 2, considering the content of the three sets $A_{contrCon,1}$, $A_{shadCon,1}$, $A_{corrCon,1}$ according to the computation of the values for the associative array conflictNo performed through Algorithm 3, the rule with the higher number of conflicts is $r_4$, as it appears four times in the conflict sets. Therefore, a query of type A is posed to the administrator, asking them if all packets satisfying the condition $c^4$ should be subject to the action $a^4$. Supposing that the administrator answers “yes” to this question, all the conflicts involving $r_4$ are solved. When the shadowing anomaly between $r_4$ and $r_{13}$ is solved, $r_{13}$ is directly removed from the cluster $g_1$. Instead, when the three correlation anomalies where $r_4$ is involved are solved (i.e., the ones with $r_5$, $r_7$ and $r_{14}$), three constraints are generated for the SMT problem: $\text{greater}(4, 3) = \text{true}$, $\text{greater}(4, 5) = \text{true}$, $\text{greater}(4, 14) = \text{true}$. After these operations, the three conflict sets have become as follows: $A_{contrCon,1} = \{\{r_6, r_{11}\}\}$, $A_{shadCon,1} = \emptyset$, $A_{corrCon,1} = \{\{r_1, r_2\}\}$, $\{\{r_1, r_{11}\}\}$, $\{\{r_1, r_{17}\}\}$, $\{\{r_2, r_{14}\}\}$.

In the second iteration of the first code block of Algorithm 2, the rule $r_1$ is selected as the subject of a query of type A, as it is involved in three conflicts. Supposing that the administrator again answers “yes”, other three correlation conflicts are solved by generating the corresponding constraints for the SMT problem: $\text{greater}(1, 2) = \text{true}$, $\text{greater}(1, 11) = \text{true}$, $\text{greater}(4, 14) = \text{true}$. After these operations, the three conflict sets have become as follows: $A_{contrCon,1} = \{\{r_6, r_{11}\}\}$, $A_{shadCon,1} = \emptyset$, $A_{corrCon,1} = \{\{r_2, r_{14}\}\}$.

In the third iteration of the first code block of Algorithm 2, each rule for which not all conflict has not yet been solved ($r_2$, $r_6$, $r_{11}$, $r_{14}$) is involved in a single conflict, i.e., $\text{conflictNo}(r_2) = \text{conflictNo}(r_6) = \text{conflictNo}(r_{11}) = \text{conflictNo}(r_{14}) = 1$. Therefore, it is not possible anymore to ask queries of type A to solve multiple conflicts simultaneously.

After this termination condition is reached for the first code block of Algorithm 2, in the second code block two queries of type B are asked to the administrator, related to the two remaining conflicts, so that the user can singularly address them. Supposing that the administrator decides that in the remaining contradiction conflict the rule that must be preserved is $r_{11}$, then $r_6$ is simply removed from the cluster $g_1$. Instead, supposing that the administrator decides that in the remaining correlation conflict the winning rule is $r_{14}$, then a constraint for the SMT problem is generated to establish its priority with respect to $r_2$: $\text{greater}(14, 2) = \text{true}$.

After Algorithm 2 ends, the cluster $g_1$ has become as follows: $g_1 = \{r_1, r_2, r_3, r_4, r_5, r_{11}, r_{14}, r_{16}\}$. Consequently, eight integer variables $x_1, x_2, ..., x_8$ are created for establishing the ordering among those eight rules. These variables are used to create an SMT problem, composed of all the constraints derived from the execution of Algorithm 2, and all the basic constraints illustrated in Section VI-C2. After creating this problem, the Z3 solver finds a correct solution based on this variable assignment: $x_1 = 4$, $x_2 = 3$, $x_3 = 5$, $x_4 = 14$, $x_5 = 1$, $x_6 = 2$, $x_7 = 11$, $x_8 = 17$. Consequently, the reordered cluster is $g_1' = \{r_4, r_3, r_5, r_{14}, r_1, r_2, r_{11}, r_{17}\}$. Indeed, there may have been other valid solutions that are compliant with all the previously described constraints. However, as all of them are valid, it is enough that the solver finds one of them.

The whole process repeats for the other two clusters, even though related anomalies are much less and easier to solve. On the one hand, for the cluster $g_2$, supposing that the administrator answers “yes” to a query of type A based on the rule $r_{16}$, both correlation conflicts are solved in a single step, and the resulting reordering operation is trivial, leading to: $g_2' = \{r_{16}, r_7, r_8\}$. On the other hand, the cluster $g_3$ is composed of a single rule, because $r_{12}$ does not have any anomalous relationship with any other rule. Therefore, no query is actually needed for that cluster.

Finally, the three reordered clusters are combined. For simplicity, we can assume that there is no specific requirement from the administrator about their relative priority, and therefore the final solution simply concatenates $g_1'$, $g_2'$ and $g_3'$. The resulting firewall policy is reported in TABLE 2. Note that Algorithm 1 should be reapplied to this policy to solve any sub-optimization that may have been caused by this reordering, but in this specific use case no new sub-optimization was actually created.

The application of the proposed methodology to this use case showcases the benefits in terms of optimization. In
particular, the solution to the anomaly resolution problem was identified through only five queries (three queries of type A and two queries of type B), which were enough to solve all fifteen anomalies. In the traditional manual approach for anomaly resolution, the human administrator should have manually solved all the anomalies one by one, with a much higher overhead in his workload. Moreover, the reordered policy is proved to be a correct solution, thanks to the correctness-by-construction principle on which the SMT problem formulation is based.

VIII. CONCLUSIONS AND FUTURE WORK
This paper presented an assisted optimized approach for firewall policy anomaly analysis and resolution. Human administrators are guided toward the resolution of all the anomalies affecting a manually written policy by answering a series of well-posed queries. The formulation of these queries is defined so as to reduce their number, thus lessening the workload of the administrators. Satisfiability checking is also employed to formulate the anomaly-free rule reordering problem as an SMT problem, thus providing assurance about its correctness. The proposed approach is convenient with respect to both manual anomaly resolution, which commonly fails in solving all anomalies or introduces other ones, and automatic strategies, which are potentially dangerous because their conflict-fixing process is not under the administrators control. These benefits, including the optimization in terms of workload reduction, have been showcased by describing how the framework implementing our proposed approach is applied to a realistic comprehensive use case.

As future work, we plan to extend the proposed models and algorithms to make them compliant with problems related to attribute-based access control, thus providing a broader generalization of the approach. We also plan to perform empirical assessments with network administrators to assess how much the proposed approach can help administrators in reducing the number of anomalies in real-world environments. Finally, we will investigate an extension of the proposed resolution strategy to inter-firewall policy anomaly, for which additional features must be accounted, such as rule matching number, firewall location in the defense-in-depth networking environment, and bi-directional configuration.

REFERENCES


**Daniele Bringhenti** received the M.Sc. degree (summa cum laude) and the Ph.D. degree (summa cum laude) in computer engineering from the Politecnico di Torino, Torino, Italy, in 2019 and 2022 respectively, where he is currently a Post-doctoral Researcher. His research interests include novel networking technologies, automatic orchestration and configuration of security functions in virtualized networks, formal verification of network security policies.

**Licia Seno** received the B.S. and M.S. degrees in automation engineering and the Ph.D. degree in information engineering from the University of Padova, Padova, Italy, in 2004, 2007, and 2011, respectively. Since then, she has been with the Institute of Electronics, Information Engineering, and Telecommunications, National Research Council of Italy, where she is currently a Researcher. Her research interests include formal verification of systems and security protocols.

**Fulvio Valenza** received the M.Sc. degree (summa cum laude) and the Ph.D. degree (summa cum laude) in computer engineering from the Politecnico di Torino, Torino, Italy, in 2013 and 2017, respectively, where he is currently a Tenure-Track Assistant Professor. His research activity focuses on network security policies, orchestration and management of network security functions in SDN/NFV-based networks, and threat modeling.