

Moving Target Defense-based Denial-of-Service mitigation in Cloud environments: A Survey

Minh Nguyen, Saptarshi Debroy

Email: mnguyen@gradcenter.cuny.edu, saptarshi.debroy@hunter.cuny.edu

Abstract—With the increased frequency and intensity of Denial-of-Service (DoS) attacks on critical cloud hosted services, resource adaptation schemes adopted by the cloud service providers (CSP) need to be intelligent. Specifically, they need to be adaptable to attack behavior and be dynamic to curb resource over-utilization. The concept of Moving Target Defense (MTD) has recently emerged as an effective and agile defense mechanism against DoS attacks that particularly target cloud hosted applications. However, the existing surveys that seek to explore this space either focus more on MTD for generic cyber attack mitigation or on DoS attack defense on cloud systems. In this survey, we particularly provide an in-depth analysis on how MTD can help recover critical cloud assets in the face of DoS attacks and how emerging programmable technologies such as, Software-Defined Networking (SDN) can be leveraged to achieve that goal. Unlike existing surveys, we categorize DoS attacks on cloud platforms based on their working mechanism. We also discuss the non-MTD based DoS defense strategies for both cloud and non-cloud infrastructures in order to highlight the pros and cons of MTD based strategies. We introduce MTD working mechanisms and present how existing research is envisioning MTD’s application in mitigating DoS attacks, both with and without SDN. We also take an in-depth look at the testbed implementations and resilience, and performance evaluations of MTD approaches. Finally, we articulate the existing challenges in MTD for DoS mitigation in cloud systems and how these challenges are shaping the future research in this domain.

Index Terms—Cloud Computing, Moving Target Defense, Denial-of-Service Attacks, Software-defined Networking.

I. INTRODUCTION AND BACKGROUND

A. Cloud versus Classical Computing

With the high demand of online services that are spatially and temporally diversified, data migration to cloud platforms have proliferated due to its cloud resources’ scalability and elasticity [1]. Before the cloud era, most of the enterprise assets in terms of services and data were stored in dedicated physical hardware - the bigger the enterprise assets, greater the need for such physical resources. However, in this solution resources do not scale well with increased load and consequently cost explodes with increased resources. Cloud computing on the other hand provides on-demand cyber resources (i.e., computing, storage, and networking) over the Internet with subscription based pay-as-you-go pricing model for its customers [1]. This enables enterprises that are consumer service or content providers to rent elastic cyber resources from public or private cloud service providers (CSPs), such as Amazon Web Services (AWS) [2], Microsoft Azure [3], Google Cloud [4], GENI [5], and CloudLab [6] instead of buying, owning, and maintaining physical data centers and servers. Consequently,

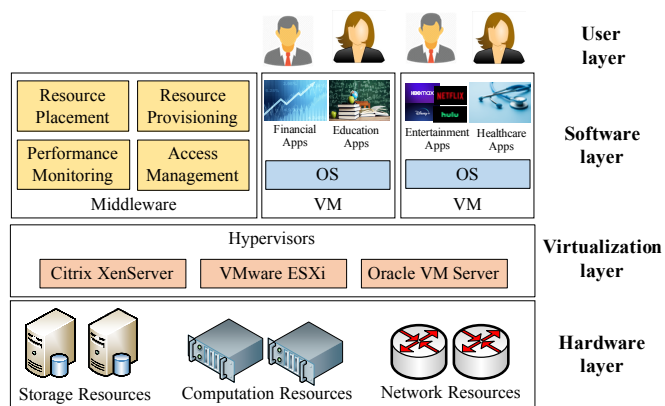


Fig. 1: Layered architecture of a cloud environment hosting critical services through provisioning distributed and elastic services

cloud has become part of the critical infrastructures for hosting essential services and data in areas such as finance, education, government services, and healthcare.

As illustrated in Figure 1, a simple abstraction of cloud infrastructure usually consists of four layers: 1) the bottom-most hardware layer provides the physical resources such as, CPUs, storage units and network resources (switches, cables), 2) the virtualization layer in the middle hosts the hypervisor (a.k.a virtual machine monitor or VMM) that creates the virtualized environment, 3) the middle layer that cross-cuts other layers by providing distributed services, such as resource management and monitoring, and 4) the software layer running the virtual machines (VMs) that host critical services/applications. The service model offered by the CSPs to the client services can also come in different packages: i) Infrastructure-as-a-service (IaaS) provides a data center and a way to host client VMs and data, ii) Platform-as-a-service (PaaS) provides a programming environment to build and manage the deployment of an application, and iii) Software-as-a-service (SaaS) provides delivery of softwares to the service users. Seamless integration among different layers of the cloud architecture and easy implementation of different service models within the same datacenter are made possible by adoption of programmable technologies, such as, software-defined networking (SDN) [7], OpenFlow [8, 9], OpenDaylight [10], and OpenStack [11].

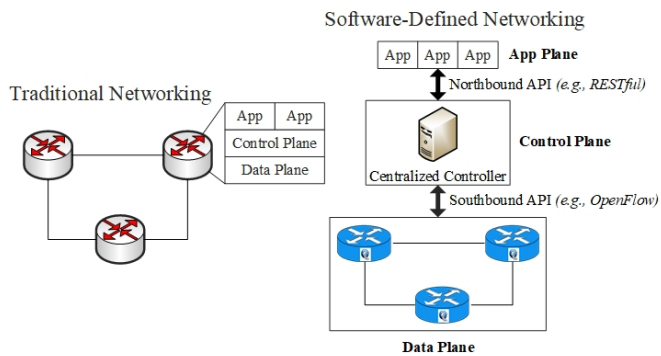


Fig. 2: Traditional Networking vs. Software-Defined Networking.

B. Adoption of SDN for Efficient Cloud Management

It can be argued that if ‘virtualization’ was the most critical technology towards realization of cloud computing, then ‘softwarization’ is the primary reason behind the ‘cloudification’ of most of the critical services. And no technology other than SDN [7, 8, 12] was more responsible behind such ‘softwarization’. SDN allows the networks to be programmable, making them more flexible and agile, i.e., network changes are made through software rather than hardware. To achieve this, SDN decouples the control plane (i.e., network control) and data plane (i.e., network functions) and enables the control plane to become directly programmable. This allows the underlying network infrastructure to be abstracted for applications and network services (in the app plane) [7] as shown in Figure 2. The control plane hosts a centralized SDN controller that acts as the brain of the network and manages flow control to data plane via Southbound APIs (e.g., OpenFlow [8, 9]). Data plane consists of the network devices (e.g., switches/routers) that follow the rules handed down from the controller and perform the corresponding forwarding functions. The interactions between the app plane and the controller use Northbound APIs (e.g., RESTful [13]). Unlike traditional networks where each network device contains the entire network stack of data plane, control plane, and app plane (as shown in Figure 2), SDN with its decoupled and centralized control is more dynamic, adaptable, and agile.

The recent advances in software defined technologies that uses OpenFlow protocol have made it easier to manage and control distributed cloud data centers across geographic boundaries. SDN has allowed public and private CSPs to implement fine-grained and dynamic network control (i.e., routing, switching, identity and access management, resource provisioning) across its data centers based on diverse vectors such as, data type and size, sources and destinations of data streams, privacy and security requirements, resource availability to name a few. The role and impact of SDN towards efficient management of distributed cloud services can not be overstated, none more so than for institutional private clouds that support data-intensive science. Fig. 3 shows an exemplar institutional private cloud infrastructure that features SDN with OpenFlow switches at strategic traffic aggregation points within the campus and backbone networks that features

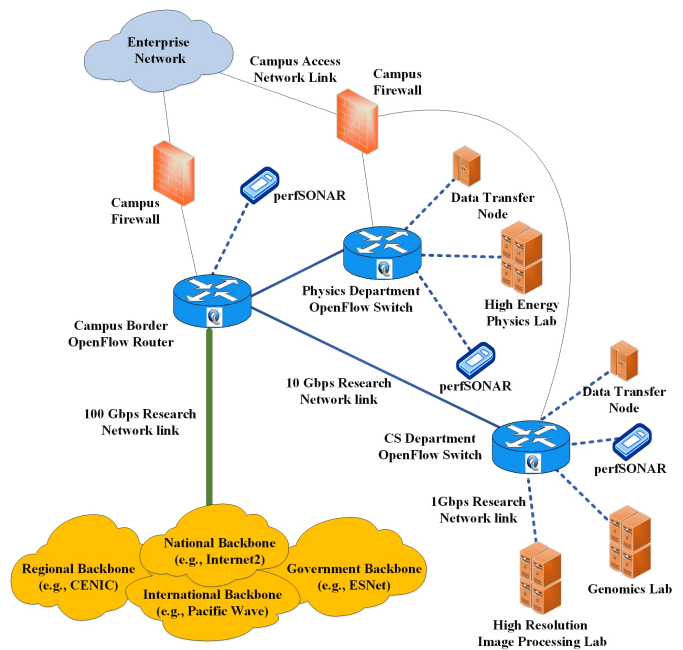


Fig. 3: An exemplar institutional private cloud with Science DMZ infrastructure enabled by SDN and OpenFlow

science DMZ (demilitarized zone or perimeter network, sits between an internal network and an external network [14]) for friction-free data-intensive science workflows [15]. SDN provides centralized control on dynamic science workflows over a distributed network architecture, and thus allows proactive/reactive provisioning and traffic engineering of flows in a unified, vendor independent manner [9]. It also enables fine-grained control of network traffic depending on the QoS requirements of the application workflows. In addition, OpenFlow enabled switches help in dynamic modification of security policies for large flows between trusted sites when helping them dynamically by-pass the campus firewall [16]. The figure also shows the infrastructural components of the institutional Science DMZ within the campus network. Normal application traffic traverses paths with intermediate campus firewalls, and reaches remote collaborator sites or public cloud sites over enterprise IP network to access common web applications. However, data-intensive science application flows from research labs that are ‘accelerated’ within science DMZs bypass the firewall to the high-speed backbones.

C. Denial-of-Service on Cloud and Mitigation Challenges

The proliferation of cloud hosted enterprise and scientific service has made the entire cloud ecosystem an enticing target of cyber attacks, in particular to Denial of Service (DoS) attacks that lead to Loss of Availability (LoA) of services through resource exhaustion. A DoS attacker typically accomplishes such exhaustion by flooding the target directly or indirectly with malicious traffic (usually with spoofed source IP addresses) till the target’s resources are overwhelmed when it cannot respond or simply crashes [17] - thus starving the legitimate users (of the cloud services) from critical services. A Distributed Denial of Service (DDoS) attack is an extreme

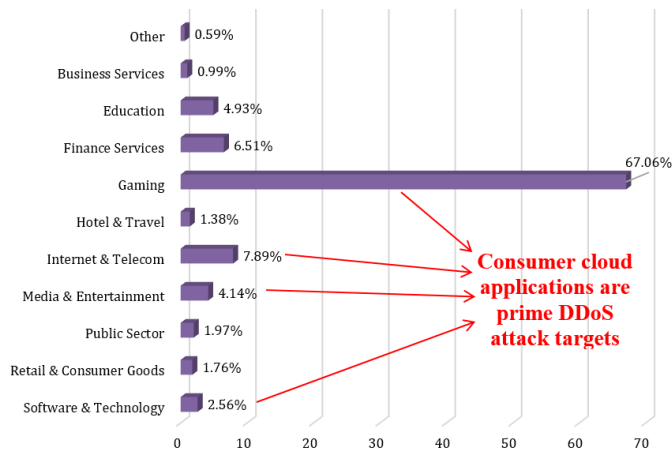


Fig. 4: Akamai Technologies 2019 survey showing DoS attack frequency by industry with more than 80% of attacks targeting consumer cloud applications.

version of DoS attack. Although conceptually DDoS attack mechanism is the same as DoS, in DDoS the attacker commands and controls an army of bots or zombies (botnets, usually contain malware-infected smartphones, personal computers, IoT devices, routers, etc.) [17, 18] that collaboratively and simultaneously bombards the target with attack intensity clocking hundreds of Gbps and Tbps. Such volume can cause even the most resilient of the systems to buckle effectively and quickly resulting in service unavailability to a large population of users. Besides, when a sudden and large surge in traffic happens at a server, it is called a flash crowd or flash event [19]. Although flash events happen with far less frequency than DoS attacks and usually caused by legitimate traffic, it still poses issues for CSPs as its characteristics are very similar to that of DoS/DDoS attacks. In fact, some attackers try to masquerade their DoS attacks as flash events [20].

In Akamai Technologies’ 2018 State of the Internet report [21], overall DoS/DDoS attacks went up (16%) in Q4 2018 in comparison to 2017 showing a steady year-to-year increase. The same report from 2019 [22] shows that DoS/DDoS attacks continue to target the cloud industry as more than 80% of the attack events have targeted cloud based consumer applications (e.g., Gaming or Internet & Telecom, etc.) as shown in Figure 4. The largest DDoS attack ever recorded happened in February 2020 when AWS cloud services saw peak attack traffic at a rate of 2.3 Tbps [23, 24]. High volume DoS attacks are not only restricted to consumer cloud applications; collaborative cloud environments such as GitHub [25] are also targets, e.g., 2018 DDoS attack on GitHub had a volume of 1.35 Tbps via 126.9 million packets per second (pps) [26]. Lack of adequate defense and recovery strategies to counter against such attacks can impact cloud service provider (CSP) reputation and cause millions of dollars in damages to cloud tenants.

The DoS attack defense challenges within a cloud platform are more severe in the following two ways. Firstly, a cloud environment becomes a vulnerability amplifier to traditional cyber security threats due to the fully distributed and highly

elastic nature of the infrastructure resources designed to serve a large population. E.g., one of the largest DDoS attacks in the history was launched on cloud based DNS servers of Dyn, Inc. [27] in 2016 (peak at 1.2 Tbps) [28] that percolated to different layers of the Internet, crippling not only the CSPs such as AWS, but also popular cloud hosted content providers such as Netflix and Twitter. The attack impacted millions of users on the east coast of the United States of America with close to 12 hours of service outage resulting loss of money and subscription [28].

Secondly, new means of attack exist that specifically target the vulnerable areas cloud environments such as application multi-tenancy, decentralized network management, and third-party broker services (between the CSP and the consumers). E.g., when hackers carry out a cyber-attack to a cloud based service, they can either try launching the attack from outside targeting the server IP address and/or the DNS; or they can infiltrate the internal network of the CSP hosting streaming services and target vulnerable virtual machines (VMs) that either have security soft-spots and/or catering to a large population of consumers for greater impact. Although such network infiltration based attacks are difficult to carry out requiring increased attack budget, most sophisticated attacks on cloud based services are network infiltration based. What makes matters worse is that such infiltration based attacks endanger the entire cloud environment, i.e., individual VMs, underlying operating systems, and hardware infrastructure by making them vulnerable to a plethora of other attacks [29].

D. Moving Target Defense (MTD)

In order to tackle the aforementioned challenges, the cloud security community and even federal organizations are exploring ‘Cyber Agility and Defensive Maneuver’ (CAADM) mechanisms [30] that are: (a) agile in response to attack detection, (b) cost effective for the CSP, and (c) sophisticated in tackling intelligent attack strategies. The goal is for such mechanisms to allow real-time service restoration through agile cloud resource adaptations once a DoS attack is detected. The same mechanisms can also limit proliferation of detected attacks within the cloud infrastructure through preventive VM resource maneuvers. Amongst the CAADM mechanisms, Moving Target Defense (MTD) based resource obfuscation/adaptation strategies are most effective to protect critical cloud-hosted applications [31]. For instance, MTD-based mechanisms are used to perform both: (i) *proactive* resource adaptation, to detect a DoS attack and act defensively before major damage is inflicted, and (ii) *reactive* resource adaptation, to act defensively after an attack has occurred. At the same time, MTD-based mechanisms are amenable to leverage the emerging Software-Defined Networking (SDN) [7, 8, 12] paradigm to achieve dynamic network resource management [32].

However, there are three distinct issues that makes the design of such MTD-based CAADM strategies non-trivial. Firstly, with every dynamic resource adaptation, the CSP encounters cost involving wastage of cloud network/compute/storage resources, which becomes especially

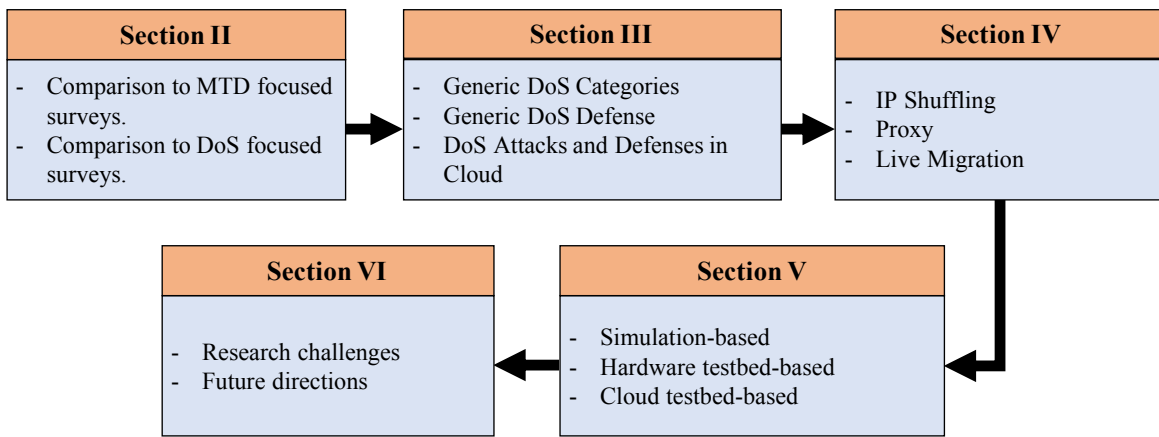


Fig. 5: Overall survey roadmap and contributions

prohibitive for proactive adaptations. However, the alternate approach of infrequent adaptations can leave the application vulnerable to DoS threats. Thus, there is a need to optimize the frequency of proactive adaptations. Secondly, with either proactive or reactive resource adaptation, the legitimate users of a cloud-hosted application will experience service interruptions and quality of experience (QoE) degradation to some extent. Such degradation can be sustained if the resource adaptations are sub-optimal and do not capitalize on the inherent heterogeneity of CSP resources to optimize performance. Thus, there is a need to optimize the CSP resource utility in the adaptations without noticeably impacting the end-user performance. Thirdly, successful MTD-based defense implementations need to possess the potential for deception, wherein a quarantine environment traps the attacker without his/her knowledge to learn more about the attack strategy, while the defense adaptations are progressing to continue service to legitimate users.

E. Contribution of this survey

These challenges have prompted cyber security community across academia, federal government, and private enterprise to explore the utility of MTD inspired attack reflection and recovery strategies in cloud environments that are efficient and yet cost effective. In recent years, researchers have also conducted extensive surveys on a decade worth of works in this space primarily targeting two focus areas: i) MTD inspired cyber defense mechanisms [33–36] and ii) DoS/DDoS defense mechanism for cloud hosted services [37–39]. Compared to these works, our survey seeks to focus more on the application of MTD strategies in clouds infrastructure against DoS attacks and how programmable technologies such as, SDN are being leveraged to implement such strategies.

In particular, the main contributions of this survey are as follows:

- Unlike existing surveys, we categorize DoS attacks on cloud platforms based on their working mechanism, e.g., volume-based, protocol-based, and application-based attacks in order to shed more light on attack process.
- Unlike other MTD focused surveys, we discuss the non-MTD based DoS defense strategies for both cloud and

non-cloud infrastructures. This gives us a better perspective on the pros and cons of MTD based strategies against DoS and DDoS.

- In this survey, we introduce MTD working mechanism and present how existing research envisioning MTD’s application in mitigating DoS attacks, both with and without SDN. We take a unique approach to categorize MTD based strategies on the basis of maneuvering techniques, e.g., IP shuffling, live migration, and proxy management. Unlike existing surveys, this approach of categorization highlights the different MTD implement strategies at various network abstractions.
- We take an in-depth look at the testbed implementations and resilience performance evaluations of MTD approaches. For example, we discuss how existing research uses cloud testbeds, hardware testbeds, simulation, and sometimes combination of these to demonstrate strategy effectiveness. We also showcase different performance (i.e., usability) and security metrics used for such demonstration.
- Finally, we articulate the existing challenges in MTD for DoS mitigation in cloud systems and how these challenges are shaping future research in this domain.

The rest of the paper is organized as follows. Section II discusses the comparison of our survey with existing surveys. Section III discusses DoS/DDoS attack classification. Section IV introduces different MTD based mitigation strategies. Section V discusses the evaluation methods and metrics. Section VI highlights the existing challenges and future directions. Section VII concludes the paper. The overall paper organization is illustrated in Figure 5.

II. COMPARISON WITH EXISTING SURVEYS

In Table I we compare the focus of our survey against recent popular surveys that primarily focus on MTD base cyber attack mitigation. In one of the early ones [33], Cai et al. conducted a thorough survey on MTD-based mitigation techniques. The authors present a function-and-movement model to provide different perspectives for understanding MTD research works. With this model, they systematically survey MTD works based on three main areas, e.g., theory, strategy, and evaluation.

Related Work	Cai et al. (2016) [33]	Zheng et al. (2019) [34]	Sengupta et al. (2020) [35]	Cho et al. (2020) [36]	Our survey
Focus on cloud and SDN	Partially	Partially	Comprehensively	Comprehensively	Comprehensively
In-depth analysis on DoS attacks	No	No	No	No	Comprehensively
Non-MTD defense methods	No	No	No	Partially	Comprehensively
MTD techniques and implementations	Partially	Comprehensively	Comprehensively	Comprehensively	Comprehensively
Evaluation methods	Comprehensively	Partially	Comprehensively	Comprehensively	Comprehensively
Future direction and challenges	Partially	Partially	Comprehensively	Comprehensively	Comprehensively

TABLE I: Comparison of key contributions of MTD.

Within these areas, the authors further classified MTD into sub-categories based on techniques and characteristics of MTD strategies. However, the survey needs a better categorization of MTD techniques and implementations keeping the most recent work in mind. It also misses some key perspectives such as, types of DoS attacks and state-of-the-art experimental testbeds for evaluation of MTD based strategies.

In 2019, Zheng et al. published an extensive survey on MTD [34] based cyber defense mechanisms. This work is focused on architectural aspects and classifications of MTD strategies. The authors categorize MTD strategies based on the level of implementation within the system stack, e.g., OS level, software/application level, and network level. For each level, the authors further categorize MTD based on techniques such as IP randomization, virtualization, and decoy among others. However, this survey is not focused on cloud systems and SDN capabilities. Furthermore, the survey lacks a comprehensive discussion on the evaluation methods and metrics for the existing MTD techniques.

Recently in [35], Sengupta et al. present an extensive survey on MTD techniques for Advanced Persistent Threat (APT) [40, 41] in SDN based cloud environments. In this survey, the authors provide an in-depth analysis on the implementation and evaluation of MTD techniques and how technologies such as SDN and Network Function Virtualization (NFV) can aid MTD implementation. The authors categorize MTD techniques based on the inter-relationship between different phases of APT. Moreover, the survey introduces a common terminology library that can help readers understand more about the underlying assumptions and threat models of existing MTD techniques. Besides, the authors conduct a thorough study on MTD evaluation techniques with various security and usability metrics. How the survey lacks DoS-focus and does not provide an extensive study on non-MTD based DoS mitigation techniques that are essential to appreciate and understand the pros and cons of MTD.

In another recent work [36], Cho et al. conducted a comprehensive survey on MTD's application for a wide range of cyberattacks in cloud, SDN, and IoT environments. They classify MTD techniques and with their respective pros and cons based on three types of operations: shuffling, diversity, and redundancy. The different types of MTD techniques are

discussed in the context of different attack vectors, e.g., shuffling, diversity, and redundancy. Besides, the authors also extensively discuss the evaluation methods and metrics (performance and security) used to validate the performance of the MTD techniques. Although similar to [35], this work is very comprehensive, it does not focus on DoS or non-MTD based works targeting DoS attacks.

Table II illustrates the comparison between our survey and other surveys that focus on DoS and DDoS based attacks and defenses in cloud environments. In [37], Yan et al. performed extensive survey about DoS/DDoS attacks in cloud infrastructures and especially in SDN environments. The authors study how DoS attacks can be launched in cloud environments and how defense mechanisms can be designed against those attacks by exploiting SDN programmability. However, the authors do not discuss the current research on state-of-the-art experiments and evaluation methods. Recently, in [38], Agrawal et al. also conducted a comprehensive survey about DoS in cloud environments. The authors classify DoS attacks based on various forms of high-rate and of low-rate attacks and discuss their strategies and impacts. Besides, they categorize the defense approaches and their performances based on multiple evaluation metrics; although they only partially analyze attacks in SDN environments. More recently, Yurekten et al. conducted a thorough survey on SDN-based defense for cyber attacks that includes DoS/DDoS attacks [39]. The authors categorize cyber attacks by examining the five-phase cyber threat intelligence. However, this survey do not focus on cloud environments. As for defense, they provide SDN-based defense mechanisms that can be used to cope with aforementioned cyber threats based on detection, prevention, and mitigation aspects. Finally, the authors discuss the evaluation techniques based on addressed threat category, defense type, defense strategy, and underlying solution approach.

Compared to these surveys, our survey focus more on clouds infrastructure and SDN aspects of the network design. We provide in-depth analysis of how MTD strategies are implemented on SDN environments to mitigate DoS attacks and how those approaches stack up against each other, both in terms of usability and security. In this survey, we specifically classify DoS attacks into three categories based on the mechanism, i.e., volume-based, protocol-based, and application-based. For the

Related Work	Yan et al. (2016) [37]	Agrawal et al. (2019) [38]	Yurekten et al. (2020) [39]	Our survey
DoS in Cloud	Comprehensively	Comprehensively	Partially	Comprehensively
DoS in SDN	Comprehensively	Partially	Comprehensively	Comprehensively
Evaluation methods	No	Comprehensively	Comprehensively	Comprehensively
Focus on MTD	No	No	No	Comprehensively

TABLE II: Comparison of existing surveys about DoS/DDoS and cyber attacks.

MTD strategies, we categorize them based on the maneuvering techniques such as, IP shuffling, proxy, and live migration. In our survey, we also present non-MTD approaches for defending against DoS in cloud environments. We also discuss the existing challenges in MTD based DoS defense and future directions to address those challenges.

III. DOS ATTACKS AND MITIGATION IN CLOUD

A. Overview of DoS attacks

Broadly, DoS attacks can be categorized into three different types: volumetric attacks, protocol attacks, and application attacks [42]. In real-world scenarios, attacks could be launched as a combination of the three in order to increase the devastating effects. In this section, we discuss these attack strategies, whereas the mitigation strategies will be discussed further in Subsection III-B.

1) *Volumetric attacks*: Volumetric attacks are classic DoS attacks where the goal is to deny service by typically creating congestion and saturation of bandwidth at the target (e.g., server) and the target network. This makes it impossible for legitimate users of the service to communicate with the server under attack. Typical examples of volume-based attacks are UDP flood, ICMP flood (a.k.a. ping flood), and amplification attacks (a.k.a. reflection attacks). In UDP flood, a large volume of UDP packets bombards a server that makes the server check for processes that are listening to the ports and respond to each UDP packet. This leads to denial-of-service for the regular clients. UDP flood as a matter of fact is behind the very first documented DDoS, the attack on University of Minnesota in July, 1999 [43–45]. Ping flood (ICMP flood) is another type of volume-based attack where the objective is to consume the victim server’s bandwidth usually by sending ICMP echo requests as fast as possible. Due to the way ICMP works (for each request, there is a reply) [46], ping flood ends up consuming the attacker’s bandwidth as well. However, there are ways to work around this feature.

A more sophisticated and potentially dangerous type of volumetric attacks are amplification attacks (a.k.a. reflection attacks) where instead of the real target, a vector is targeted that can reflect and amplify the attack traffic towards the real target. Typical example is the DNS amplification attack where the attacker makes large number of requests to DNS (Domain Name System) [47] servers with spoofed source IP addresses and the destination is changed to the target’s IP address. As a result, the DNS servers forward the volume of responses to the victim. Other popular amplification attacks include NTP (Network Time Protocol) [48] and SSDP (Simple Service Discovery Protocol) [49] amplification. In these, the

attackers typically exploit the bad design of a UDP-based request and response protocol (e.g., DNS or NTP) and trigger a significantly larger number of responses than the original amount. Although the recent discoveries of new amplification vectors are rare, once such attacks hit the target, more often than not they lead devastating consequences. For example, the previously mentioned recent DoS attacks on AWS [24] and GitHub [26] were both amplification attacks that exploited rather newly found vulnerabilities of CLDAP (Connection-less Lightweight Directory Access Protocol) [50] and Memcached [51] protocols.

2) *Protocol attacks*: Protocol attacks exploit weaknesses in the working mechanisms of Network and Transport layer protocols to cause a service denial by over-consuming the server resources and other equipments in the network infrastructure (e.g., firewalls and load balancers). Typical examples of protocol attacks include SYN flood and IP fragmentation attacks. One of the first recorded DoS events on the Internet in 1996 was a SYN flood attack [52] followed by many other high profile DoS attacks in the history [52–54]. Typically TCP (Transmission Control Protocol) [55] uses a three-way handshake to establish a connection: 1) the client sends a SYN message to the server to request a connection; 2) the server acknowledges the request by sending SYN+ACK message back to the client and leave an open port waiting for the final acknowledgment; and 3) the client responds with the final ACK message and the connection established. In SYN flood, the attacker exploits the last feature by not only not sending back the final ACK packet, but also sending more SYN packets leading to denial-of-service at the server for other legitimate users due to the lack of ports.

Another example of protocol attacks is IP fragmentation attack which exploits the network Maximum Transmission Unit (MTU) [56]. IP fragmentation process mandates that any transmitted IP packets larger than the network MTU (e.g., 1500 bytes for Ethernet [57]) will be broken into IP fragments which will later be reassembled at the final destination [58]. The attacker exploits this mechanism by preventing the packets to reassemble at destination (e.g., by only sending a part of the packet), resulting in service unavailability. Other protocol attacks includes Ping of Death and Smurf that are exploit ICMP [59]. However, they are largely considered solved for contemporary hardware/software systems [59].

3) *Application attacks*: The final broad category are application attacks where the attackers seek to exhaust the target server’s resources by exploiting the vulnerabilities of network applications (e.g., web servers). In general, application attacks usually are considered most sophisticated and mitigation tech-

niques are rather complex. Typical examples of application attacks include HTTP flood and low-and-slow attacks. In HTTP flood, the attacker floods the target web server with HTTP GET packets (used to requests for images, files, etc. from a server) and/or HTTP POST packets (used to send data to a server and/or database in order to create/update a resource) [60]. This consumes not only bandwidth, but also disk space and available memory of the target server. Beside the previously mentioned attack in 2018 [26], GitHub had also suffered the largest DDoS attack ever at the time in 2015 which was a HTTP flood attack [23]. Another popular application attacks are low-and-slow attacks. Low-and-slow attacks operate by requesting the targeted server to execute some tasks and then sending data to the server at a very slow pace in order to keep the tasks unfinished for long time. As a consequence, the server has to always keep the connection open in order to finish the requested tasks and which in turn denies other tasks from other legitimate users. Examples include attacks using tools such as Slowloris [61, 62] or ‘R.U.D.Y.’ a.k.a. ‘R U Dead Yet?’ [63].

B. Generic DoS defense and mitigation

In operational settings, several traditional DoS defense strategies have been adopted for generic, non-cloud networks in order to minimize impacts of volumetric and protocol DoS attacks. For example, usage of firewalls and filtering can help mitigate DoS attacks by dropping malicious traffic and control what traffic can reach the infrastructure. However, firewalls can also lead to false positives, i.e., filtering out legitimate packets. Moreover, firewalls can be susceptible to high volume flood attacks since firewalls’ state tables can only hold a certain number of sessions. For some particular attacks, disabling or limiting some functionalities can help prevent DoS. For example, disabling UDP support by default is a good method to cope with Memcached amplification attack or reducing the number of open DNS resolvers to can help limit DNS amplification attack. Intelligent routing and diversion techniques such as using Content Distribution Networks (CDNs) [64] or Load Balancers [65] can help breaking the massive traffic into manageable chunks as well as preventing direct traffic to important parts of your system. Although this approach is very effective, it requires a lot of resources and therefore may not suitable for resource constrained environments. When it comes to application attacks, due to their complexity, defenders usually have to combine multiple defense methods such as firewalls, pattern adaptation, and even incoming requests rate limiting in order to be effective [66, 67]. Nevertheless, in many cases these traditional operational methods fall short; thus, new intelligent approaches have been proposed.

One such approach is packet payload intervention at servers or routers [68–71]. Server side intervention of SYN cookies [68] is a popular method to fight SYN flood attack. Here, upon sending SYN-ACK packet back to the client the server drops the original SYN request from the queue. If the ACK message eventually arrives, the server rebuilds the SYN packet using a cryptographic technique. Consequently, there always remains available ports for new handshake establishments and thus new

connection are not denied. For router side intervention, work such as, [69, 70] try to manipulate the essential information inside the packet payload to come up with defense strategies. In [69], the authors propose router stamping that helps identify the source the DoS attacks hidden with IP spoofing [54]. If a packet travels via three routers, each router will record the IP address of its predecessor before it forwards the packet. By counting the number of stamped packets at each router during an attack, the routers can anticipate the source of the attack. In [70], the authors propose DoS defense viz., NetFence at bottleneck routers i.e., the routers at service provider side that are responsible for inbound traffic. Bottleneck routers stamp the packets that carry congestion monitoring feedback to signal congestion to access routers; while other access routers use it to monitor senders’ traffic. These congestion monitoring feedbacks are encrypted so that they cannot be faked.

Another way of payload modification for DoS defense is implementation of Pushback that was first presented in [71]. Pushback method considers a DDoS attack as a congestion problem by dropping the traffic at the congested points and propagating the information back to upstream routers in order to force them to rate-limit the traffic i.e., Pushback. The authors also propose a heuristic algorithm to filter bad traffic that further improves Pushback mechanism. Nevertheless, the payload modification approach has some limitations. Firstly, it requires cooperation between router manufacturing companies such as, Cisco [72] and Juniper [73], or software platforms such as, Linux Foundation [74] or FreeBSD Project [75] in order to make the modified packets compatible with router hardware and drivers. Secondly, this approach has to monitor and/or modify the packet payload which could add significant overhead on packet encapsulation and decapsulation process and certain inaccuracies (e.g., SYN cookies).

Another intelligent approach viz., Honeybot [76–78] has been very popular in mitigating DoS. Honeybot is a decoy system designed to act like a real system with data and resources having no legitimate use [76, 77]. It is typically is set outside the internal network in order to lure attackers to perceive it as the real system. Most often, honeybot is configured as part of the most external layer of the network or the science DMZ [14]. That way, if the network is under attack, honeybot will be hit first. For sophisticated attacks, it is possible that both the honeybot and the network/server are attacked simultaneously [76]. However the honeybot can help in quick analytics on the attack traffic and use that information for recovery and/or quarantine [76, 78]. In works such as [78] honeypots are even used to identify the infrastructures behind DNS amplification attacks. Although effective, honeypots are not designed to mitigate DoS attacks; rather to act as decoys for analytics and information collection. Furthermore, if not properly deployed, honeybot could attract unwanted attack traffic which could lead attackers to penetrate the internal networks [76]. Although these methods can be applied to cloud environments, due their bigger scale and existence of new threats, cloud environment inspired new approaches towards DoS defense.

C. Traditional DoS defense in cloud environments

DoS attacks targeting cloud services and infrastructure also fall under the aforementioned three categories, i.e., volumetric attacks, protocol attacks, and application attacks. For DoS defense strategies in cloud, based on the focus of this survey we categorize them into two categories: Traditional or non-MTD based and MTD-based. The MTD-based defense strategies will be discussed in Chapter IV. Here we introduce the traditional or non-MTD approaches for DoS defense in cloud infrastructure. Broadly many DoS defense strategies applied to non-cloud infrastructures can be borrowed for cloud environments. However, many authors have proposed new methods that leverage the uniqueness of cloud infrastructure such as sofwarization, virtualization, elastic (e.g., on-demand), etc. For DoS/DDoS defense and mitigation designed for cloud infrastructures, work such as [79–92] are notable that can be broadly categorized into groups shown in Figure 6.

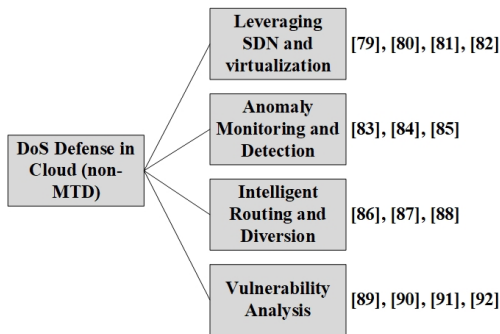


Fig. 6: Categories of traditional DoS defense strategies in cloud infrastructures

1) *Leveraging SDN and virtualization*: These group of works [79–82] use the programmability and virtualization of SDN enabled cloud infrastructure to defend against DoS/DDoS attacks. Authors in [79] (Salah et al.) implement a cloud-based overlay network (i.e., a virtual network built on top of physical networks) that provides an integrated set of on-demand security services such as, Intrusion Detection Systems (IDS), DDoS prevention, and firewalls. In [80], Fayaz et al. present Bohatei, a flexible and elastic system that leverages SDN and virtualization with a resource management algorithm to drive malicious traffic through the defense system while minimizing latency and network congestion. In [81], Zhang et al. propose Poseidon, a volumetric DDoS defense strategy that is adaptable to attack patterns and leverages SDN’s programmable switches combining the advantages of hardware-based and software-based defenses. Similarly, Liu et al. in [82] take advantage of SDN’s programmable switches to introduce Jaqen, a programmable switch-native tool that can run detection and mitigation functions without relying on additional hardware.

2) *Anomaly Monitoring and Detection*: This group of works [83–85] propose anomaly monitoring and detection strategies to segregate anomalous traffic. Narayana et al. in [83] propose a SDN-based path query language for efficient path-based traffic monitoring that can help measuring the flow of traffic, which is crucial for many tasks, including DoS

mitigation. Elsabagh et al. in [84] propose Cogo, a proactive probabilistic system for early detection and mitigation of application DoS attacks such as low-and-slow attacks. In [85], Demoulin et al. introduce FineLame, a framework for detecting asymmetric DoS attacks (attacks that target applications’ internal algorithms or semantics) via resources monitoring.

3) *Intelligent Routing and Diversion*: These works [86–88] propose intelligent routing and subsequent diversion techniques to isolate attack traffic. Works such as, [86] (Gilad et al.) leverage the flexibility of cloud resources to deploy an affordable CDN-based solution namely CDN-on-Demand to mitigate volumetric DoS attack and flash crowds. In [87], Ramanathan et al. propose SENSS, a security service that can help the victims to ask the upstream Internet Service Providers (ISP) for help by requesting on-demand attack monitoring and filtering. Authors in [88] (Zheng et al.) propose DynaShield, a on-demand low-cost crypto-based solution that can auto-scale to large attacks to cope with volumetric DDoS attacks.

4) *Vulnerability Analysis*: Unlike other works that propose solutions based on existing DoS vulnerabilities, this final group of works [89–92] investigates new threats and vulnerabilities and measures weaknesses and strengths of cloud-based solutions using case studies. In [89], Vissers et al. investigate attack vectors in which attackers can exploit to discover the IP addresses of important parts inside the infrastructure of cloud-based security providers and also evaluate their impacts. In [90], Bushart et al. present DNS unchained, a new amplified application-based DoS attack against DNS authoritative servers and its impacts. In [91], Jansen et al. investigate a volumetric DoS case study against Tor anonymity network [93] via some default Tor bridges that reside on popular CSPs. Kopp et al. in [92] investigate the impact and anatomy of booter-based DDoS. Booters are DDoS-as-a-service providers that offer their customers DDoS services for an affordable price.

It is important to notice that most of these works focus on volumetric and application attacks, while there exists very few novel work on protocol attacks. This is quite understandable because to perform a successful protocol attack, the attackers first have to discover a network or transport layer protocol vulnerability and then exploit that. Since there are limited number of de-facto and standardized network and transport protocols on the Internet and most of their vulnerabilities have been discovered and patched, there is not much left to exploit. Further proof for this is that most protocol attacks such as SYN flood, Ping of Death, Smurf, and part of IP fragmented attacks have been considered largely solved with newer software updates. However, the advent of SDN and OpenFlow has significantly changed the landscape. Although it opens a broader scope of DoS defense in cloud, SDN and OpenFlow also come with their own vulnerabilities [94–96]. One of these new vulnerabilities in OpenFlow protocol have been exploited to launch a reflection-based attack, viz., table-miss [97–101] that can completely cripple both the switches and the controller.

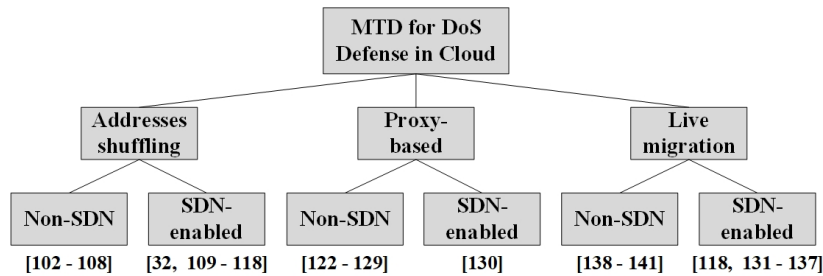


Fig. 7: MTD for DoS defense in cloud environments.

IV. MTD FOR DOS MITIGATION IN CLOUD

In order to address the rapid growth of DDoS attacks, the cloud security community and federal organizations are exploring ‘Cyber Agility and Defensive Maneuver’ (CAADM) mechanisms for cloud that can allow for real time service restoration through agile cloud resource adaptation once an attack is detected, and also limit proliferation of detected attacks within the cloud environment through preventive maneuvers [30]. In order to realize such CAADM mechanism in cloud, MTD-based techniques are the need of the hour [31], as: 1) Intelligent but fast converging algorithms can be developed for both proactive and reactive maneuvers based on triggers for a wide range of global and local greedy optimization criteria; 2) Emerging network management technologies such as, SDN can help implement and operationalize such dynamic and agile maneuvers in order to evade impending attacks; and 3) Sophisticated dynamic maneuvers can be designed to create system obfuscation helping to deceive/illude the adversary in a false sense of success and thus stopping the proliferation. In this survey, we broadly categorize the current research landscape in MTD based DoS defense for locud environments into the following three categories based on the adopted MTD based maneuvering mechanism, viz., network addresses shuffling based, proxy based, and live migration based. Then for each category, we further sub-categorize the works into the following two groups based on the adoption of SDN or other programmable technologies for MTD implementation, viz., non-SDN MTD and SDN-enabled MTD. The overall classification is illustrated in Figure 7. Below we discuss the theoretical and system design details for each such category. The evaluation techniques adopted for each such work and corresponding results are later discussed in Section V.

A. Network addresses shuffling

Network addresses shuffling and randomization is the classic approach and one of the most popular implementations of MTD in cloud. In this technique, network addresses (e.g., IP addresses) associated with the application servers or virtual machines (VMs) are reassigned or randomized around an available pool of addresses (e.g., from DNS Servers) periodically (can be fixed or adaptive). The cloud service users who are oblivious to such randomization are then redirected to new IP addresses without significant quality of service (QoS) drop (Figure 8). Such randomization considerably increases attacker cost it has to continuously guess the network addresses or

address space associated with the target server or VM. In recent times, SDN-enabled shuffling and randomization work are gaining momentum. In most cases such implementations are reactive in nature, i.e., the defense scheme kicks in once an attack is detected. However, SDN can also be useful in implementing such maneuvering proactively thanks to the complete centralization of the network control plane and can help prevent impending attacks. The usage of the decoupled SDN controller allows easily deployment of monitoring, predictive, and defensive algorithms that work in complete harmony. Among the works that employ network addresses shuffling, [102–109] are notable for non-SDN based methods, while works such as, [32, 110–117] propose SDN-enabled shuffling and randomization techniques.

1) *Non-SDN implementations:* Among these works, Carroll et al. in [102] present probabilistic models for IP addresses shuffling based MTD. These models quantify the attacker success under different conditions such as network size, number of addresses, and number of vulnerable systems. The authors investigate the relationship between shuffling frequency and connection loss and found that shuffling provides limited protection against attackers focusing on one high-value system. Their results also indicate that shuffling is acceptable if there is a small pool of vulnerable systems within a large network address space, but may cost connection losses of legitimate users. In [103, 104], Alavizadeh et al. investigate the effectiveness of individual shuffle, diversity, and redundancy based MTD techniques and propose a method that combines all three. They use a graphical security model, viz., Hierarchical Attack Representation Model (HARM) [118] to model and analyze the MTD techniques. Wang et al. in [105] study the MTD timing problem, i.e., the optimal time to conduct the adaptations and to balance the cost-effectiveness. The authors devise a multi-module framework along with a cost-effective adaptation algorithm called Renewal Reward Theory-based solution (RRT) to cope with this issue.

In [106, 107], Clark et al. present a game-theoretic framework that combines decoy network and address space randomization to distract and mislead adversaries. The proposed framework consists of two components: i) one that differentiates between the decoy nodes interactions with the adversary and a real node and ii) another for adversarial game formulation in order to find the attack target in a network consisting of real and decoy nodes. Moreover, the authors argue that the designed framework only needs to randomize IP addresses if the adversarial scanning rate exceeds a certain threshold. Nizzi

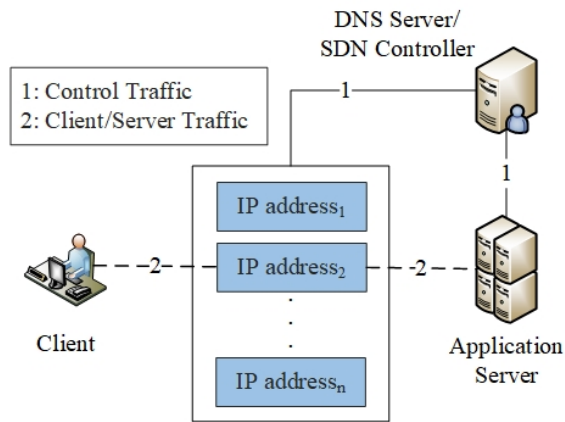


Fig. 8: Logical diagram of MTD implementation through IP Shuffling

et al. in [108] present a cryptography based address shuffling algorithm namely ASHA for IoT devices in wireless sensor networks (WSN) to deal with security threats including DoS attacks. The proposed algorithm uses address renewal methods by leveraging cryptographic hash functions that aims to be simple, collision-free, and low overhead. Likewise, in [109], Yao et al. also propose a network address shuffling approach for IoT devices to eliminate security threats in WSN. The authors formulate the problem as a stochastic cost optimization problem and propose a novel stochastic cost minimization mechanism (SCMM) to solve it.

2) *SDN-enabled implementation*: Among these works, Kampanakis et al. in [110] analyze how SDN can be used for MTD by investigating the advantages and disadvantages of network-based MTD techniques. The authors argue that programmability with SDN controller can help with system/resource adaptations which is an important factor to MTD maneuverability. Also, a highly programmable SDN based system can provide obfuscations that will increase the cost of an attack by making the attacker spend more resources in order to study the attack surface(s). Steinberger et al. in [111] investigate how MTD can leverage SDN to the fullest extent. The authors argue that if MTD strategies are implemented using SDN in a collaborative environment, the impact of large-scale DDoS attack can be significantly reduced. They argue that MTD can limit the attacker's knowledge of the target due to the ever-changing attack surface (because of MTD) and thus can increase attacker cost. The advantage of their collaborative DDoS defense solution is that using their system, each participating partner achieves insights into the current threat landscape. Further, collaborative DDoS defense pools expertise and resources from all collaborating partners, thus achieving greater success against attacks. This work also indicates that ONOS [119] is an appropriate SDN OS to enforce implementation of MTD due to its guaranteed scalability as ONOS has been used and tested in several high-speed networks. In [112], Zhou et al. propose a new cost-effective shuffling (CES) method against DDoS attacks using MTD based on game theory. CES takes shuffling frequency in to account and model the interaction between the attacker and

defender using Multi-Objective Markov Decision Processes. Based on this model, the authors study the best trade-off between the effectiveness and cost of shuffling in each particular scenario.

Jafarian et al. in [113, 114] propose an address randomization technique called Random Host-address Mutation (RHM) to mitigate reconnaissance attacks. This technique can turn the servers into untraceable moving targets by leveraging SDN to mutate their original network addresses. The actual IP addresses (rIP) are kept unchanged but it can create routable short-lived ephemeral IP addresses (eIP) from the unused ranges of the network address. The eIP addresses are provided via DNS and ARE used for routing. They are automatically translated back into the rIPs and vice versa at the network edges close to the destination. RHM utilizes a two-level mutation scheme to maximize the unpredictability: i) low frequency mutation (LFM) that changes the set of unused ranges assigned to each host and ii) high frequency mutation (HFM) that assigns the new eIP address associated with each host.

In [32] and [115], Chowdhary et al. seek to tackle DDoS attacks by selecting suitable countermeasure based on obtained information about the adversaries. However, the authors choose two different paths to obtain the needed information. Work in [32] presents an automated dynamic system reconfiguration by leveraging scalable Attack Graphs (AG) to assess the attacks and select necessary countermeasures to perform real-time network reconfiguration, both proactively and reactively. A node in an AG is a combination of hosts and the possible vulnerabilities that exist on that particular host. Each host may have intra-connections or inter-connections with other hosts. Hence, if a botnet communicates with clients to target a system resource, this information can be modeled and tracked. This scheme also ensures that there is no security policy violation or conflict after the adjustments are done. Whereas in [115], the authors combine SDN-enabled MTD with Intrusion Detection System (IDS) to formulate a threat scoring system based on vulnerabilities and IDS alerts and select MTD countermeasure. This defense mechanism is called MASON and instead of IP addresses, it uses port hopping technique. Based on threat scores, MASON can identify network services with high-security risk and takes corresponding actions.

Aydeger et al. in [116] present a signaling game to thwart the emerging Crossfire attack, a type of Stealthy Link Flooding Attack (SLFA) attack. It is a variant of DDoS attacks that congests the connections surrounding the network of the target servers by sending low-volume traffic from many bots. The proposed signaling game considers the defender and the attacker as two players and the equilibria represents the best strategies for each player. Based on the game results, the authors propose an improvement upon Random Route Mutation (RRM) [120], viz., Strategic RRM. It is a multipath routing algorithm that periodically changes routing to avoid passing through some compromised links or nodes [121]. Similarly, Xu et al. in [121] also propose an improvement over route mutation algorithm for MTD. They model route mutation process as a Markov Decision Process and introduce a Context-aware Q-learning RM algorithm (CQ-RM) that can

learn attack strategies to optimize the selection of mutated routes adaptively.

In [117], Nguyen et al. propose Whack-a-Mole, a SDN-driven MTD mechanism for DDoS defense in cloud environments. Whack-a-Mole resource maneuvering works at two levels: i) it proactively spawns replicas of VMs hosting critical applications where the applications are seamlessly migrated and ii) it mutates the IP addresses associated with the services by assigning the VM replicas with IP addresses belonging to different address spaces (assuming that the entire cloud network is divided into different address spaces). Upon resource maneuver, the OpenFlow switches with the help of SDN controller direct all new incoming user requests to the spawned VMs whereas the existing users are allowed to finish their sessions with the old VMs. Upon completion of the existing users' sessions, the VMs are terminated and IP addresses that will be recycled for newly spawned VMs. In their work, the address mutation is optimized to keep the new IP address selection as unpredictable as possible to increase attacker cost.

B. Proxy-based

In this method, the IP address of the real server or VM (which in most cases is the target) is concealed from all clients and the real servers hide behind a group of intermediate proxy machines or VMs. Clients first communicate with the control unit (e.g., authentication server) that directs them to the correct proxy. The MTD based maneuvering is initiated periodically (fixed or dynamic) when the physical and/or logical identity of the proxy that is being connected to the real server is changed to another (as shown in Figure 9). The identity of the new proxy can be random or based on some intelligent mechanism. Thus, for the attacker trying to target a server or VM, figuring out the identity of the proxy is essential and for obvious reasons non-trivial. Unlike other categories of MTD works, most of the current state-of-the-art proxy-based MTD techniques such as, [122–129], except [130] can be implemented successfully in spite of not having SDN like programmability in the system.

Jia et al. in [122] and Wang et al. in [123] propose MOTAG, a proxy-based MTD mechanism that utilizes a layer of secret moving proxies to mediate all communications between the clients and the protected VMs. The filters deployed surrounding the VMs only allow traffic from the valid proxy nodes. The proxy system act as a shield between the VMs and the rest of the Internet. When one proxy node is under attack, it is replaced by another node at a different network location and the associated clients are redirected through the new proxy. With the proposed algorithms, the proxy nodes can also be used as an isolated environment for the potentially malicious users working as insiders. Similarly, Wood et al. in [124] devise a relay network called DoSE that act as a proxy between clients and servers. The relay node is located in the public cloud infrastructure and content delivery networks (CDN) help to disseminate the relay information to the corresponding clients. DOSE aims to achieve low cost DDoS attack mitigation for small to medium-sized organizations that typically

have limited budgets. DOSE connects clients to relay proxies and proposes new methods for assigning clients to relays in order to mitigate network layer attacks while minimizing costs.

Fleck et al. in [125] and Kesidis et al. in [126] extend the work on MOTAG and utilize it to proactively minimizing DDoS attack's impact by attempting to thwart potential attacks during the reconnaissance phase. The authors study a proactive and cloud-side MOTAG defense in which proxies dynamically change to thwart DDoS attack's reconnaissance phase and consequently reduce the attack's impact. In these works, the authors use a load balancer to direct clients to the proxies. They use an adversarial coupon collection based mathematical model to formulate the problem. In [127], Bandi et al. also present a MOTAG-like strategy combined with Fast-Flux, a technique used to hide the servers behind an ever-changing system of proxy. The authors propose FastMove, a shuffling algorithm to determine the number of legitimate clients on each proxy server in order to save the largest possible number of clients.

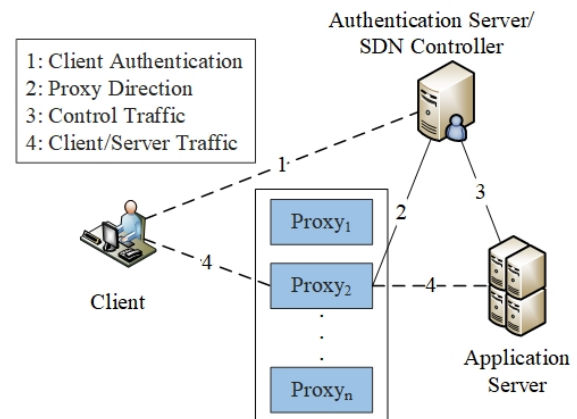


Fig. 9: Logical diagram of MTD that implements proxy-based mechanism

In [128], Venkatesan et al. argue that proxy-based defense mechanisms such as MOTAG and DoSE can be vulnerable to a new type of attack, namely the proxy harvesting attack. The proxy harvesting attack exploits a weakness in the authentication process of these proxy-based architectures to collect information about a possibly large number of proxy nodes with the help of insiders. To overcome the proxy harvesting attack, the authors propose BIND-SPLIT strategy that limits the number of IP addresses that can be harvested; combined with the proactive defense mechanism called PROTAG. The proposed PROTAG mechanism help with two primary factors: i) proxy selection to determine the optimal proxies to be replaced and ii) movement frequency to determine the optimal time to replace the proxies.

In [129], Wright et al. introduce a novel game theoretic model that formulates a DDoS attack as a two-player normal-form game between the attacker and defender. In this game, both sides want to affect the quality of experience (QoE) of the legitimate clients, while keeping their own costs low. This work is called MOTAG Game as it is built upon MOTAG model with an objective to evaluate the effectiveness of proxy-based MTD strategies. To achieve that, the authors use the

simulation-based empirical game-theoretic analysis (EGTA) to find game-theoretic equilibria in complicated games over restricted strategy spaces.

As mentioned earlier, most of the proxy-based MTD techniques are implemented without having SDN like programmability in the system, except [130]. Here Aydeger et al. propose a Shadow Network (SN) framework to deal with Crossfire attack (this work shares the same goals as in [116]). SNs are tiny low-cost networks (can be virtual or physical) attached to actual ISP network are used to deceive attackers with the fake topology information. In this proposed framework, the ISP network is assumed to be SDN-based through which the ISP can perform traffic-engineering to any traffic flow using the SDN Controller. The location of SNs are similar to proxies as they reside between the protected servers/VMs and the clients. Nevertheless, the SNs do not completely shield the entire internal network like the proxies but rather a part of it.

C. Live migration

In this method, a pool of VMs working as application servers (these VMs can be created beforehand or on the go) are responsible for hosting the target services. In order to create system obfuscation, periodically (can be fixed or dynamic) the services are migrated from one set of VMs to another by a control server as shown in Figure 10. The VMs currently not hosting any services are kept on standby and ready to host at a moment's notice. Such migrations are often called live migrations as all the migration related actions such as, taking a VM snapshot and transferring the files from one VM to another happen 'online' while the users are still using the services. The optimal VM or set of VMs selection for migration can be based on complex algorithms that consider factors such as, VM's capacity, current specifications, network address space where it resides among many other. After the migration is complete, the users of the service are redirected to the new VM(s). This considerably increases the attacker cost as the attacker has to first identify the VMs currently hosting the target services before launching an attack. In many of the current research [117, 131–133], the live migration and user redirection are often carried by the centralized SDN controller. The decoupled and centralized SDN controller provides added flexibility and dynamicity to apply a host of intelligent algorithms that are effective against impending attacks. However, there also exists a number of research works [134–136] that achieve VM migration using traditional non-SDN based methods.

1) *Non-SDN implementation*: Among the works that propose traditional non-SDN based migration, Jia et al. in [134] propose a cloud-enabled, shuffling-based MTD strategy to marginalize the attackers within a space of VMs. When under attack, it replicates the attacked VM instances, migrates it to a newly instantiated replica VM at a different network locations, and assigns legitimate clients to the newborn VM. In order to prevent moving sophisticated attackers to the new replica VMs, the authors keep track of the legitimate client assignments and if the new replica is attacked, they separate benign client sessions from potentially malicious ones. Through multiple rounds of shuffling, they can filter out

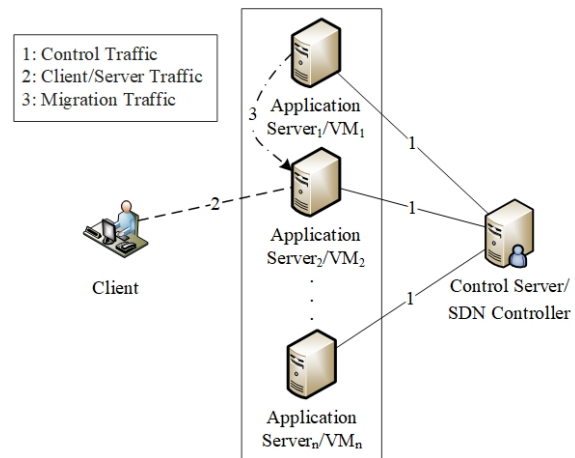


Fig. 10: Logical diagram of MTD implementing live VM migration and user redirection

the attackers and enclose them. The authors also introduce a novel family of algorithms to optimize the mitigation runtime and minimize the number of shuffles.

Peng et al. in [135] models a cloud-based system with heterogeneous resources and dynamic attack surfaces to ascertain whether and to what extent MTD is effective. The authors use a VM migration technique that moves the snapshots of servers within the pool of VMs. The novelty of this work is its consideration of the attacker's accumulated knowledge about the attack surface and formulation of a stochastic problem that wants to minimize the probability of the service being compromised.

In [136], Venkatesan et al. propose a MTD approach to defend against stealthy botnets for resource-constrained environments. The authors deploy detectors across the network and apply a series of defense strategies to periodically change the placement of those detectors. The objective is to make attackers uncertain about the location of detectors so that they have to perform additional actions in an attempt to create detector-free paths through the network; hence, increasing the attackers' likelihood of detection.

2) *SDN-enabled implementation*: Among the works that implement VM migration using SDN, Debroy et al. in [131, 132] propose a DDoS defense mechanism that allows for proactive migration of target application for impending attacks and triggers reactive migration when under attack. This work's novelty is in the moving frequency optimization and the ideal location selection for migration across heterogeneous pool of VMs based on attack probability. The objectives for such optimization is to make the migration frequent enough to evade impending attacks at the same time not too frequent that it causes unnecessary resource wastage. In order to find the ideal VM, the authors propose an optimal market-driven approach that is based upon distributed optimization principles. This approach uses virtual market economics in order to optimize resource allocation during migration. Besides, as part of the reactive defense, the authors also present false reality scheme that reuses the attacked VM as a trap to deceive the attacker and gather adversarial information.

Approaches	Features	Performance against Volumetric Attacks	Performance against Protocol Attacks	Performance against Application Attacks
Network Address Shuffling [32, 102-118]	<ul style="list-style-type: none"> - A pool of network addresses (e.g., IP addresses) is managed dynamically. - Network addresses of the target server(s) are reassigned or randomized periodically. - Requires less physical resources. 	<ul style="list-style-type: none"> - Works for most volumetric attacks because the targets are masqueraded under different network addresses. - Not effective against DNS amplification attacks as IP addresses are resolved during attacks. 	<ul style="list-style-type: none"> - Works for most protocol attacks because the targets are masqueraded under different network addresses. 	<ul style="list-style-type: none"> - Does not work for application attacks due to the fact that such attacks target domain names rather than network addresses.
Proxy-based [122-130]	<ul style="list-style-type: none"> - The target server(s)' identities are concealed from all clients behind a group of intermediate proxies. - The identities of proxies can be static or dynamic. - Requires more compute and network resources for the pool of proxy servers. - For effective implementation, may need support from other tools such as firewalls. - Proxies can also help in early detection of attacks. - Using proxies can open up other vulnerabilities [128]. 	<ul style="list-style-type: none"> - Works for volumetric attacks as the targets are protected behind proxies. - The proxies act as the first line of defense and face the brunt of the attack 	<ul style="list-style-type: none"> - Works for flooding based protocol attacks (e.g., SYN flood) as the targets are protected behind proxies. - Does not work for stealthy protocol attacks (e.g., IP fragmentation) that can percolate through the proxies, even when they are SDN enabled. 	<ul style="list-style-type: none"> - Works for flooding based application attacks (e.g., HTTP flood) as the targets are protected behind proxies. - Does not work for low and slow protocol attacks that can percolate through the proxies, even when they are SDN enabled.
Live Migration [118, 131-141]	<ul style="list-style-type: none"> - A pool of physical/virtual resources are kept in the standby to host target services. - The services are migrated to and from these resources. - Such migrations can be proactive (i.e., periodic) or reactive (i.e., when under attack). - Such redundancy typically requires resource abundance. - Often requires SDN based implementation with other strategies (e.g., proxy-based) to be effective. 	<ul style="list-style-type: none"> - Works for volumetric attacks as the targets are moved around proactively using a SDN controller. - Works even when there is only a reactive scheme (with or without SDN) where the target can be quickly migrated to safety. 	<ul style="list-style-type: none"> - Works for flooding based protocol attacks (e.g., SYN flood) as the targets are moved around with or without SDN. - Works for stealthy protocol attacks (e.g., IP fragmentation) as long as the attacks are detected early and SDN migrates the target(s) rapidly. 	<ul style="list-style-type: none"> - Works for flooding based application attacks (e.g., HTTP flood) as the targets are moved around with or without SDN. - Works for low and slow protocol attacks as long as the attacks are detected early and SDN migrates the target(s) rapidly.

TABLE III: Summary of different MTD strategies and their relative utility against common DoS attacks

In [133], Nguyen et al. propose a Common Vulnerability Scoring System (CVSS) [137] driven Bayesian Attack Graph (BAG) model designed for low-budget and small-scaled private cloud infrastructures such as Campus Private Clouds (CPC). This model is used to perform a dynamic threat/risk assessment for integrity, confidentiality, and availability attacks on data residing in the VMs. BAGs are used to model cyber attack causal relationship and used for assessing attack success likelihoods. The likelihood of an attack success is calculated using CVSS. As for the vulnerabilities, the authors use relevant cyber-attack statistical data from Common Vulnerability and Exposures (CVE) [138]. Using the proposed model, they perform a case study on the campus network to evaluate the likelihood of success of confidentiality, integrity, and availability attacks with and without MTD based maneuvering. Finally, in Whack-a-Mole [117], Nguyen et al. propose a live VM spawning and migration scheme where replicas of VMs hosting critical applications are proactively spawned and where the cloud-hosted applications are migrated to. Once a VM is spawned, all the new service requests routed to the newly spawned VM whereas the old VM only services the existing users. Each VM has a lifetime after which its resources are reclaimed. This lifetime is optimized to serve the average span of a service request session in order to minimize user QoE degradation. At the same time, the authors optimize the spawning frequency that is large enough to thwart an impending attack, yet not too large that it causes unnecessary resource wastage and from too frequent spawning.

A comparative summary of the aforementioned different MTD strategies and their relative pros and cons against common types of DoS attacks (discussed in Section III) is described in Table III.

V. EVALUATION METHODOLOGIES

In this section, we discuss and compare the evaluation methodologies used in MTD based DoS defense for cloud systems discussed in Section IV. In order to broadly capture the different types of evaluation strategies used in such works, we categorize them into three groups: simulation-based, hardware testbed-based, and cloud testbed-based as shown in Table IV. For each of these groups, we discuss how the related works evaluate the most important metrics to gauge the proposed strategies' success, viz., security and usability. As shown in Fig. 11, security metrics are typically evaluated using attack probability, attack graphs, and risk computation. Whereas, most important and relevant usability metrics to be evaluated are cost and performance. Fig. 11 also illustrates the list of works corresponding to each such metric.

Groups	Simulation-based	Hardware-based	Cloud-based
Works	[102–104, 108]		
	[106, 107, 109]	[105, 110–112]	[125, 128, 134]
	[121–123, 126]	[113–116]	[131, 132]
	[124, 128, 129]	[127, 130]	[117, 133]
	[117, 131–136, 139]		

TABLE IV: Categories of evaluation methodologies used in MTD based DoS defense works

A. Simulation-based

In the absence of hardware testbeds and cloud testbeds, simulation-based (sometimes numerical) approach is a good first step towards evaluating the success of the proposed strategies. In this method, authors typically use software platforms such as MATLAB [140] and available datasets to simulate their methods and models. Due to the lack of real experiment setups, authors mostly use this method in analysis work and rarely use it alone to evaluate system performance. Simulation results are typically used along with the other two evaluation methods. Works such as, [102–104, 106–109, 117, 121–124, 126, 128, 129, 131–136, 139], use extensive simulation results to demonstrate system performance.

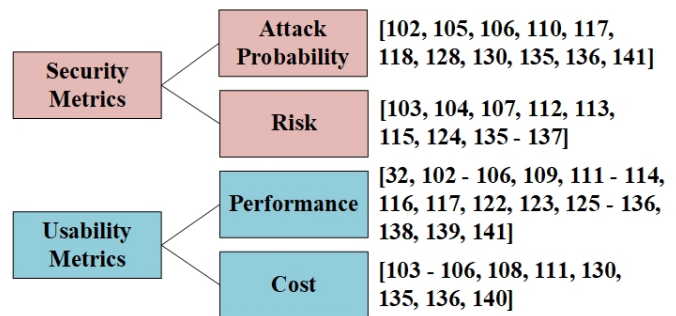


Fig. 11: Security and usability metrics used in related works

1) *Security metrics*: For attack probability approaches, authors in [102] construct a probability model to measure the mean time to security failure where the security failure is defined by the system state being compromised by an attacker and where the system is defended by MTD based maneuvering. In works such as, [117, 131, 132], the authors simulate their Poisson point process-based model to measure the optimal moving interval to minimize the probability of getting hit by the attacks. They simulate the model for different attack budgets in terms of the ratio of attack time and idle time. Authors in [113, 114] develop probabilistic models to measure attack probability when a set of reconnaissance defenses that includes deception technique and network address shuffling as MTD are deployed in a given system, while varying the network size, the size of VMs deployment, and the number of vulnerable nodes. The authors set up a virtual network using Mininet [141, 142] and a SDN controller. The evaluation show that RHM provides a robust performance in countering sophisticated threat models for both proactive and adaptive schemes with low overhead. In [108], authors provide probabilistic models to measure the effectiveness of an address shuffling-based MTD technique with respect to the network size, the address space scanned, the degree of system vulnerability, and the frequency of shuffling operations. The results indicate that for a typical personal area network (PAN), AShA can effectively mitigate DoS attacks with tunable overhead.

Among works employing attack graph approaches, [133] simulate all combinations of an attack graph based on BAG model. They compare the attack success rate for confidentiality, integrity, and availability attacks for both MTD-based and non-MTD based approaches. In [136], authors proposed

a technique to capture the dynamic changes in the network resulting from deploying MTD during the entire simulation runtime. Using two metrics, viz., minimum detection probability and attacker’s uncertainty, the simulation results show that the proposed approach can effectively reduce the likelihood of successful attacks. Work such as, [103, 104] use the hierarchical attack representation to model a system’s security features with two layers, an upper layer and a lower layer. The upper layer represents a network’s reachability information (i.e., network topological information) while the lower layer represents a node’s vulnerability information using attack graphs. The results show that the proposed combined techniques can satisfy the evaluation criteria while individual techniques do not. Authors in [32] also analyze the main advantages of using attack graphs, viz., ease of evaluation and representation. Furthermore, attack graphs can be adopted to compute various security metrics based on the MTD application.

Among works employing risk assessment approaches, authors in [106, 107] provide new metrics for MTD evaluation and risk analysis. They propose statistical metrics to study the effect of how the attacker can quickly conduct and succeed in adversarial attacks. The authors assume that the system will always have a running task that can be measured. The results show that networks should consist of a mixture of high-interaction (that implements the full protocol) and low-interaction (that implements a subset of protocol states) VMs. In [110], authors consider game theoretical formulation of MTD systems; specifically, they model it as a Markov Game. The authors provided a theorem, subject to probabilistic constraints, to calculate the revenue for the defensive and offensive approaches in MTD systems. Their work depends on testing different defensive and offensive strategies and is tested using a networking setup that includes vulnerable services and a firewall component.

2) *Usability metrics*: Among these, [115, 121, 139] use some form of cost function for the evaluation. Authors in [121] aimed to identify an optimal interval of VM migration in order to maximize security with minimum cost based on a game theoretic formulation called Vickrey-Clarke-Groves (VCG) mechanism. The simulation results show that proposed mechanism provides significant improvements in multiple aspects including defense, mutation overhead, network, and convergence performance compared to state-of-the-art methods. In [139], authors evaluate the system cost caused by an attacker at different stages of the network. The attacker-defender game is modeled as a finite zero-sum matrix game with a bounded cost function with a mixed-strategy Saddle Point Equilibrium (SPE). Players utilize cost-function learned online to update MTD strategies. The numerical results show that feedback mechanism allows network defense to respond to unexpected events.

Among works that evaluate performance as a usability metric, authors in [113, 114] identify the virtual IP (vIP) mutation, range allocation, and range distribution constraints in order to minimize the quality of service (QoS) impact induced by vIP collisions, as well as to maintain optimal level of unpredictability. Probabilistic performance analysis of

MTD reconnaissance defense is conducted in [139]. This work analyzes quantifiable MTD metrics such as reconnaissance, deception performance, attack success probability verses connection drop probability and attacker’s success probability under different conditions such as, network-size, number of vulnerable computers. Authors in [115] use mission and attack metrics for analyzing the effectiveness of network defense. They analyze dynamic defenses such as ‘Active Re-positioning in Cyberspace for Synchronized Evasion and Self-shielding Dynamic Network Architecture’ using mission and adversary activity set. Here mission success, i.e., the rate at which mission tasks are completed and mission productivity, i.e., how often are mission tasks successful are used as QoS measurement metrics for evaluations. In [135], via simulation, the authors identify the conditions and extent of the proposed strategy’s effectiveness. From the results they conclude: i) VM migration is more effective when the pool of VMs is dense and/or when the attack is large scale and ii) the heterogeneity and dynamic of attack surface help improve the scheme’s effectiveness.

B. Hardware testbed-based

Hardware testbed-based evaluation helps verify the performance of MTD based techniques under more realistic system environments, using an actual testbed within a lab setting. However, such evaluations do not always scale well for cloud-scale systems. Regardless, many MTD based DoS defense works use small or mid-sized hardware testbeds to study and assess the performance of their proposed MTD based techniques. Among these, [105, 110–116, 127, 130] are notable.

1) *Security metrics*: Among these works that evaluate security metrics, work such as [105, 110, 111, 116] use attack probability as their measure for security. In [105], authors use probabilistic models to measure the effectiveness of the proposed IP-multiplexing based network shuffling techniques in terms of attack probability and defense cost. Their experimental setup is created on their lab Intel Xeon server where the results indicate that the proposed framework and algorithm can provide the same security results as known methods but is more cost-effective. While authors in [110] use probabilistic models to verify the effectiveness of a port hopping-based MTD technique against reconnaissance attacks. In [111], authors consider resource availability as an important metric for analyzing impact of MTD countermeasure. The system reconfiguration rate is modeled as a function of system resources using Continuous Time Markov Chain (CTMC). The analysis of the effect of reconfiguration on the availability is considered for fine-tuning MTD decision. Work such as, [116] use a probabilistic model by building a stochastic model to describe an integrated defense system consisting of MTD based maneuvering, deception, and an IDS. They analyze the performance of the integrated defense system compared to a system with various combinations of defense mechanisms. The hardware testbed is setup using Mininet and Floodlight controller [143]. The evaluation shows that the proposed scheme can minimize the impact of attacks similar to original RRM, while it brings significantly less overhead.

Among the works with attack graph based security evaluation approaches, authors in [111] argue that an attack graph can be easily visualized and can help the network administrators identify the vulnerabilities of the network and choose appropriate defensive strategies such as, MTD. In [112], authors propose a SDN-based route mutation technique to deal with DDoS attacks that is validated via a Mininet [141] implementation with a Floodlight SDN controller [143]. Further, they define a route mutation MTD technique for the ISP network context through NFV and virtual shadows network aiming to thwart possible DDoS attack. They demonstrate that their route mutation method makes it difficult for the attackers to perform attack reconnaissance phase and obtain network topology information.

Among risk evaluation approaches, works such as, [113, 114] conduct a substantial analysis to evaluate the effectiveness of two MTD techniques that combine address shuffling and resource diversity. They consider three key metrics for security evaluation, viz., Risk (Risk), Attack Cost (AC), and Return on Attack (RoA). They demonstrate that MTD decreases Risk and RoA while increasing AC. The authors also show that combining shuffling and diversity can optimally meet these multiple objectives, whereas a single solution with either shuffling or diversity cannot. In [113], authors use an anti-coordination game to capture the interplay of choice, diversity, and scalability of risk in SDN-based MTD. This study evaluates a scenario where one node in a network is compromised while the others use a game theoretic approach to decide whether to switch or not. They extend their work in [114] by investigating eight security metrics to evaluate the effectiveness of combined shuffling and diversity. Another work [129] consider a statistical approach to evaluate the likelihood of a successful attack as risk. The authors proposed an approach to determine the minimum effort required from a system to detect stealthy botnets. The entropy was measured to determine how close an adversary is to the detection point, where high entropy indicates the attacker is far from the detector in terms of network distance. The authors use physical servers to create a SDN network consisting of VMs, SDN switches, and OpenDaylight controller [10]. The evaluation results indicate that the proposed mechanism can mitigate DDoS attacks while outperforming other existing algorithms in term of required CPU overhead.

2) *Usability metrics*: Among the works that evaluate usability on a hardware testbed, [112, 115, 129] measure usability in terms of cost. Authors in [112] present a cost effective MTD solution against DDoS and Covert Channel attacks. Through MTD adaptation, their work aims to answer two main questions: 1) what is the adaptation cost?, and 2) what is the cost incurred by a defender if an attacker succeeds in exploiting a particular vulnerability?. The adaptation cost includes any cost related to purchasing required software or hardware helping in the adaptation process. Their solution does not rely on IDS-generated alerts while making the adaptation. In [115], authors utilize change-point analysis method for MTD cost-benefit analysis for a multi-layer network resource graph. The proposed method analyzes mission productivity and attack success productivity on dynamic network address translation

(DNAT). The evaluation results show reduced attack success probability using DNAT over a network under observation. The path enumeration mechanism used in this research work can, however, suffer from scalability challenges because of frequent path probability calculation and update operations. In [129], the authors develop MASON, a periodic VM migration scheme based on the balance between the level of security obtained and the cost incurred upon the migration of VMs. The experiments are setup on a real Science DMZ testbed that consists of VMs, OpenFlow switches, and OpenDaylight controller. The evaluation results show that MASON can effectively thwart DDoS attacks. The results also indicate that situational awareness based on static vulnerability information and dynamic threat events should be used for taking MTD decisions, especially for large-scaled cloud networks.

Among the works that measure usability in terms of performance, authors in [115, 127] conduct a statistical analysis of static vs. dynamic attacks against different MTD strategies: uniform, random, diversity-based, evolution-based, and optimal. Experimental results on performance versus adaptability show that diversity-based MTD is the optimal strategy against most attack scenarios. Authors in [130] model performance parameters such as availability, downtime, and downtime cost using a Continuous Time Markov Chain model. The experimental results show that cost-effective VM migration can be performed in a SDN-based network with limited impact on network performance. The research work utilizes normalized CVSS score as a key metric for initiating VM migration.

C. Cloud testbed-based

Cloud testbed-based evaluations are probably the most widely used validation method due to the wide availability of community cloud testbeds such as, AWS [2], GENI [5], CloudLab [6], DeterLab [144], Chameleon Cloud [145], and PlanetLab [146]. Cloud testbeds provide high level of programmability to implement diverse types of attacks within a controlled environment as well as the ability to implement restriction-free and easily parameterized defense strategies at cloud-scale. At the same time, cloud scale implementation allow researchers to gain meaningful insights from ‘in-the-wild’ experiments before implementation in a real system. Finally, results obtained through cloud testbeds are easily reproducible and thus are widely accepted. Therefore, a wide range of works [117, 125, 128, 131–134] in MTD based DoS defense for cloud environments choose cloud-testbed based evaluations to demonstrate system effectiveness.

1) *Security metrics*: Among works evaluating attack probability, authors in [125] model the security of a configuration as inversely proportional to the probability with which an adversary can come up with a new attack given the attacks it performed in the earlier time steps. Their evaluation is conducted in AWS based cloud testbed for different case studies with distributed probing to demonstrate the success of the attackers in identifying the number of VMs. Authors in [128] propose a game-theoretic strategy as a deception technique for MTD to prevent remote OS fingerprinting attacks. They setup their experiments on AWS testbed to show that their proposed

technique can significantly decrease the fingerprinting attack success probability while the overall usability of the system is preserved without performance degradation. In [132], the testbed is developed in GENI with VMs, OpenFlow switches, and SDN controller. The authors use response time and average packet dropped to evaluate their reactive scheme. Besides, attack success rate is used to measure the performance of the proactive scheme. The evaluation indicates that proactive scheme successfully performs migrations that protect the target applications from DDoS attacks with a very low attack success rate, while reactive scheme can effectively mitigate DDoS attacks. Results also show that the false reality scheme successfully tricks an attacker with a false sense of success without substantially increasing the overall CSP cost.

For attack graph based security evaluation, Bayesian attack graphs have been used by authors in [133, 134] for defending the network against vulnerability exploitation attempts. In [134], the defender's problem is formulated as a Partially Observable Markov Decision Process and the optimal defense policy for selecting countermeasures is identified as a solution. The experimental environment is setup in their private cloud testbed and the results show that the proposed mechanism can save 80% of legitimate clients for a DDoS attack of 100K bots. Authors in [133] show that the security analysis of a large-scale cloud network in real time is a challenging problem. Here, attack graphs help in identification of possible attack scenarios that can lead to exploitation of vulnerabilities in the cloud network. The testbed is build in GENI cloud that simulates a campus network with campus private cloud. The experiment results show that the utility of VM live migration-based MTD strategy in successfully minimizing the attack impact and future attack success probability.

2) *Usability metrics*: The cost and effectiveness evaluation of reactive and proactive network defense strategies IS conducted by work such as [128] using Measurement of Effectiveness metrics. This work considers hop-delay for different attack success rates, and static defense policies. They show that an attacker's productivity, i.e., how quickly attacker can perform adversarial tasks increases against static defense; whereas attacker's confidentiality, i.e., ability to remain undetected is same for both the static and the dynamic defense case.

For evaluating performance, authors in [125] try to solve a multi-faceted problem where the MTD tries to obfuscate the network topology to an attacker and, at the same time, ensures that it does not negatively impact a defender's ability to debug network issues. This is done by leveraging the knowledge asymmetry about the network topology that a defender and an attacker has. Authors in [134] analyze the performance impact of placing IDS at all possible enforcement points in a cloud network. It is noteworthy that the placement of more than 15 detection agents in their simulated network fails to provide any additional intrusion detection benefit, whereas the network throughput decreases drastically from 16 Gbps in the case of a single detection agent to 6 Gbps when 15 detection agents are placed.

VI. RESEARCH CHALLENGES AND FUTURE DIRECTIONS

Here, we discuss the open challenges and future directions in MTD for cloud DoS defense research domain:

1) *More fine-grained research on proactive MTD*: Effective proactive or preventive MTD strategies can be designed in order to evade DoS attacks before they hit their target. With the emerging of programmable technology such as SDN, network can be designed where effective anomaly detection can trigger MTD if and when an impending DoS attacks if suspected. However, any false positive detection would cause considerable resource wastage from MTD related resource maneuvering which can add up quickly especially if the system is resource constrained. At the same time too infrequent maneuvering can leave the resources vulnerable to attacks and thus eventual service performance degradation. Thus the fundamental questions to address for effective and efficient MTD design for cloud infrastructure are: i) What is the optimal frequency of proactive MTD related resource maneuvering that protects the system without consuming excessive cloud resources? and ii) How to ensure such frequent proactive maneuvering does not affect the performance of the cloud hosted services?

2) *Strong coupling between MTD and Intrusion Detection/Prevention Systems (IDS/IPS)*: Most of the state-of-the-art IDS/IPS research do not include a recovery plan, especially in SDN based systems where DoS attacks can be more sophisticated to detect and prevent (e.g. table-miss attack). There are some siloed IDS/IPS work in cloud that solely detect and prevent DoS attacks based on Artificial Intelligence/Machine Learning (AI/ML) [147]. However, very few of these provide strategies where the effects of the attack can be minimized and/or cloud assets are moved to safety. Thus, there is a need for holistic approaches of IDS/IPS and MTD based recovery/evasion techniques. With the help of a customized IDS/IPS, more intelligent MTD strategies can be designed for early detection of sophisticated attack signatures and consequent early evasion and recovery.

3) *Lack of AI/ML for MTD*: In recent times, AI/ML have evolved as powerful tools towards defending against cyber attacks and privacy preservation. Although most existing MTD strategies assuming some stochastic attack behavior, this might not always be true. Thus, AI/ML integration with MTD is an obvious extension where more effective evasion and recovery strategies can be designed based on robust learning and without preconceived assumptions. Although AI/ML have been used for IDS/IPS in works such as, [147], their integration with MTD strategies is still lacking. Therefore, more research is needed towards AI/ML-driven MTD strategy design. However, research is needed to tackle the typical AI/ML challenges such as, training latency and requirement of huge datasets in order for such integration to be effective.

4) *DoS vulnerabilities for broader SD-ecosystem*: As mentioned before, the research space of DoS attacks and defense on cyber systems is not new and in recent times more focus is given on DoS vulnerabilities in cloud systems and cloud hosted services. Consequently, exploration of DoS vulnerabilities in SDN systems has also gained momentum as most cloud systems are SDN enabled. However, software-defined

ecosystems extend far beyond SDN based cloud data centers. Some examples of such frontier research spaces include SD-RAN (Software-defined Radio Access Network), SD-WAN (Software-defined Wide Area Network), SDX (Software-defined Internet eXchange points), and SDx (Software-defined everything environments) to name a few. Being software-defined, such ecosystems suffer from the same vulnerabilities as SDN from DoS attacks among many other which are specific to the use cases supported by these ecosystems. Thus, there is a need for more dedicated research on such new frontiers in broader SD-ecosystem.

5) *Lack of accessible DoS in cloud dataset for researchers:* Another pandemic in cyber attack and defense research space is the lack of state-of-the-art datasets available to researchers. Typically network attack datasets belonging to Internet service providers (ISPs) are shared and curated through facilities such as, CAIDA [148], IMPACT [149], and Kaggle [150]. The ISPs are incentivized to share such data. However, that is not quite true for cloud ecosystems as CSPs such as Google, Amazon, and Microsoft are reluctant to share DoS attack datasets in the fear of disclosing their secret sauce in terms of network design and propitiatory protocols. This is quite detrimental to the entire cloud security research community. Thus, there is a need to incentivize the CSP community (maybe through brokering by the federal agencies) in order to ensure more collaboration and cooperation between industry and academia around access to datasets.

6) *MTD for private and community cloud systems:* Finally, we argue that most of the MTD based defense strategies consider SDN based public cloud ecosystem under the control of corporations such as Amazon and Google. However, very little MTD based defense research is being done for private and community cloud platforms such as, institutional cloud facilities and high performance computing centers (HPC). These private and community clouds in many cases lack the state-of-the-art cyber defense tools and facilities as: a) they are less visible to the rest of the Internet and consequently are relatively less attractive or lucrative targets of sophisticated cyber attacks and b) they have overall operating budget constraints and lack resource redundancy. Thus many of such private and community clouds are ill-equipped to handle sophisticated DoS attacks if and when they occur [133]. Therefore, there is a need to explore more cost effective, simpler to implement, and more proactive MTD based defense strategies that do not rely on resource redundancy.

VII. DISCUSSION AND CONCLUSIONS

In this survey, we extensively studied recent notable works that explore how MTD can protect cloud infrastructures. We offered a novel categorization of MTD approaches based on maneuvering techniques such as, IP shuffling, live migration, and proxy systems. We classified DoS attacks based on their properties, e.g., volumetric attacks, protocol attacks, and application attacks. Besides, we studied non-MTD methods and DoS defense approaches for non-cloud like environments. Unlike existing surveys, we extensively discussed the role of SDN in implementing effective MTD based techniques.

We also examined various evaluation methodologies for MTD based mitigation techniques and provided our perspectives on open challenges and future directions in this space. The discussions of this survey will aid cyber security domain scientists - beginners and experts alike, cloud service providers, and network administrators in comprehensively understanding the state-of-the-art in this space and start to explore the open challenges.

REFERENCES

- [1] P. Mell and T. Grance, "NIST Special Publication 800-145: The NIST Definition of Cloud Computing," *National Institute of Standards and Technology (NIST) - US Department of Commerce*, 2011.
- [2] "Amazon Web Services (AWS)," <https://aws.amazon.com/>.
- [3] "Microsoft Azure," <https://azure.microsoft.com/>.
- [4] "Google Cloud," <https://cloud.google.com/>.
- [5] "GENI (Global Environment for Network Innovations)," <https://www.geni.net/>.
- [6] "CloudLab," <https://cloudlab.us/>.
- [7] Open Networking Foundation (ONF), "Software-Defined Networking (SDN) Definition," <https://opennetworking.org/sdn-definition/>.
- [8] —, "ONF TR-535: ONF SDN Evolution," 2016.
- [9] N. McKeown, T. Anderson, H. Balakrishnan, G. Parulkar, L. Peterson, J. Rexford, S. Shenker, and J. Turner, "OpenFlow: Enabling Innovation in Campus Networks," *ACM SIGCOMM Computer Communication Review (CCR)*, 2008.
- [10] "OpenDaylight (ODL)," <https://www.opendaylight.org/>.
- [11] "OpenStack," <https://www.openstack.org/>.
- [12] K. Yap, T. Huang, B. Dodson, M. S. Lam, and N. McKeown, "Towards Software-Friendly Networks," *1st ACM Asia-Pacific Workshop on Systems (APSys)*, 2010.
- [13] R. Fielding, "Architectural Styles and the Design of Network-based Software Architectures," *Doctoral Dissertation - University of California, Irvine (UCI)*, 2000.
- [14] US Computer Emergency Readiness Team (CERT), "Recommended Practice: Improving Industrial Control System Cybersecurity with Defense-in-Depth Strategies," *Cybersecurity and Infrastructure Security Agency (CISA) - US Department of Homeland Security (DHS)*, 2016.
- [15] E. Dart, L. Rotman, B. Tierney, M. Hester, and J. Zurawski, "The Science DMZ: A Network Design Pattern for Data-Intensive Science," *IEEE/ACM International Conference for High Performance Computing, Networking, Storage, and Analysis (SC)*, 2013.
- [16] I. Monga, E. Pouyoul, and C. Guok, "Software-Defined Networking for Big Data Science," *ACM/IEEE International Conference for High Performance Computing, Networking, Storage, and Analysis (SC)*, 2012.
- [17] United States Computer Emergency Readiness Team (CERT), "CERT Security Tip ST04-015: Understanding Denial-of-Service Attacks," *Cybersecurity and Infrastructure Security Agency (CISA) - US Department of Homeland Security (DHS)*, 2019.
- [18] M. Donner, "Phagocytes in Cyberspace," *IEEE Symposium on Security and Privacy (S&P)*, 2010.
- [19] J. Jung, B. Krishnamurthy, and M. Rabinovich, "Flash Crowds and Denial of Service Attacks: Characterization and Implications for CDNs and Websites," *ACM 11th International Conference on World Wide Web (WWW)*, 2002.
- [20] S. Kandula, D. Katabi, M. Jacob, and A. Berger, "Botz-4-sale: Surviving Organized DDoS Attacks that Mimic Flash Crowds," *2nd USENIX Symposium on Networked Systems Design and Implementation (NSDI)*, 2005.
- [21] "Summer 2018 State of the Internet - Security Report," *Akamai Technologies*, 2018.

- [22] “State of the Internet - Security: A Year in Review,” *Akamai Technologies*, 2019.
- [23] Cloudflare, “Famous DDoS attacks: The largest DDoS attacks of all time,” <https://www.cloudflare.com/learning/ddos/famous-ddos-attacks/>.
- [24] M. Pinho, “AWS Shield Threat Landscape Report is Now Available,” <https://aws.amazon.com/blogs/security/aws-shield-threat-landscape-report-now-available/>, 2020.
- [25] “GitHub,” <https://github.com/>.
- [26] S. Kottler, “February 28th DDoS Incident Report,” <https://github.blog/2018-03-01-ddos-incident-report/>.
- [27] “Dyn,” <https://oci.dyn.com/>.
- [28] “Update Regarding DDoS Event Against Dyn Managed DNS on October 21, 2016,” <https://www.dynstatus.com/incidents/5r9mppc1kb77>, 2016.
- [29] S. Yu, Y. Tian, S. Guo, and D. O. Wu, “Can We Beat DDoS Attacks in Clouds?” *IEEE Transactions on Parallel and Distributed Systems (TPDS)*, 2014.
- [30] “DARPA-BAA-15-56: Extreme DDoS Defense (XD3) - Amendment 2,” *Defense Advanced Research Projects Agency (DARPA)*, 2015.
- [31] L. M. Marvel, S. Brown, I. Neamtiu, R. Harang, D. Harman, and B. Henz, “A Framework to Evaluate Cyber Agility,” *IEEE Military Communications Conference (MILCOM)*, 2015.
- [32] A. Chowdhary, S. Pisharody, and D. Huang, “SDN Based Scalable MTD Solution in Cloud Network,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2016.
- [33] G. Cai, B. Wang, W. Hu, and T. Wang, “Moving Target Defense: State of the Art and Characteristics,” *Frontiers of Information Technology & Electronic Engineering*, 2016.
- [34] J. Zheng and A. S. Namin, “A Survey on the Moving Target Defense Strategies: An Architectural Perspective,” *Springer Journal of Computer Science and Technology*, 2019.
- [35] S. Sengupta, A. Chowdhary, A. Sabur, A. Alshamrani, D. Huang, and S. Kambhampati, “A Survey of Moving Target Defenses for Network Security,” *IEEE Communications Surveys & Tutorials*, 2020.
- [36] J. H. Cho, D. P. Sharma, H. Alavizadeh, S. Yoon, N. Ben-Ahser, T. J. Moore, D. Kim, H. Kim, and F. F. Nelson, “Toward Proactive, Adaptive Defense: A Survey on Moving Target Defense,” *IEEE Communications Surveys & Tutorials*, 2020.
- [37] Q. Yan, F. R. Yu, Q. Gong, and J. Li, “Software-Defined Networking (SDN) and Distributed Denial of Service (DDoS) Attacks in Cloud Computing Environments: A Survey, Some Research Issues, and Challenges,” *IEEE Communications Surveys & Tutorials*, 2016.
- [38] N. Agrawal and S. Tapaswi, “Defense Mechanisms Against DDoS Attacks in a Cloud Computing Environment: State-of-the-Art and Research Challenges,” *IEEE Communications Surveys & Tutorials*, 2019.
- [39] O. Yurekten and M. Demirci, “SDN-based cyber defense: A survey,” *Elsevier Future Generation Computer Systems (FGCS)*, 2020.
- [40] E. Baize, “Developing Secure Products in the Age of Advanced Persistent Threats,” *IEEE Symposium on Security and Privacy (S&P)*, 2012.
- [41] Kaspersky, “What Is an Advanced Persistent Threat (APT)?” <https://www.kaspersky.com/resource-center/definitions/advanced-persistent-threats>.
- [42] Cloudflare, “What is a DDoS Attack?” <https://www.cloudflare.com/learning/ddos/what-is-a-ddos-attack/>.
- [43] E. Osterweil, A. Stavrou, and L. Zhang, “20 Years of DDoS: A Call to Action,” *arXiv:1904.02739*, 2019.
- [44] MIT Technology Review, “The first DDoS attack was 20 years ago. This is what we’ve learned since.” <https://www.technologyreview.com/2019/04/18/103186/the-first-ddos-attack-was-20-years-ago-this-is-what-weve-learned-since/>, 2019.
- [45] US Computer Emergency Readiness Team (CERT), “CERT Incident Note IN-99-07,” *Cybersecurity and Infrastructure Security Agency (CISA) - US Department of Homeland Security (DHS)*, 1999.
- [46] J. Postel, “RFC 792: Internet Control Message Protocol,” *ISOC Request for Comments (RFC)*, 1981.
- [47] P. Mockapetris, “RFC 1035: Domain Names - Implementation and Specification,” *ISOC Request for Comments (RFC)*, 1987.
- [48] D. Mills, J. Martin, J. Burbank, and W. Kasch, “RFC 5905: Network Time Protocol Version 4: Protocol and Algorithms Specification,” *ISOC Request for Comments (RFC)*, 2010.
- [49] Y. Y. Goland, T. Cai, P. Leach, Y. Gu, and S. Albright, “Internet Draft: Simple Service Discovery Protocol 1.0,” *ISOC Internet Draft*, 1999.
- [50] A. Young, “RFC 1798: Connection-less Lightweight X.500 Directory Access Protocol,” *ISOC Request for Comments (RFC)*, 1995.
- [51] “Memcached,” <https://memcached.org/>.
- [52] C. Patrikakis, M. Masikos, and O. Zouraraki, “Distributed Denial of Service Attacks,” *The Internet Protocol Journal (IPJ) - Cisco Systems*, 2004.
- [53] C. L. Schuba, I. V. Krsul, M. G. Kuhn, E. H. Spafford, A. Sundaram, and D. Zamboni, “Analysis of a Denial of Service Attack on TCP,” *IEEE Symposium on Security and Privacy (S&P)*, 1997.
- [54] US Computer Emergency Readiness Team (CERT), “CERT Advisory CA-1996-21: TCP SYN Flooding and IP Spoofing Attacks,” *Cybersecurity and Infrastructure Security Agency (CISA) - US Department of Homeland Security (DHS)*, 1996.
- [55] J. Postel, “RFC 793: Transmission Control Protocol,” *ISOC Request for Comments (RFC)*, 1981.
- [56] J. Mogul and S. Deering, “RFC 1191: Path MTU Discovery,” *ISOC Request for Comments (RFC)*, 1990.
- [57] Microsoft, “The Default MTU Sizes for Different Network Topologies,” <https://support.microsoft.com/en-us/topic/the-default-mtu-sizes-for-different-network-topologies-b25262c5-d90f-456d-7647-e09192eeef4>.
- [58] J. Postel, “RFC 791: Internet Protocol,” *ISOC Request for Comments (RFC)*, 1981.
- [59] R. van den Berg and P. Dibowitz, “Over-Zealous Security Administrators Are Breaking the Internet,” *16th Systems Administration Conference (LISA)*, 2002.
- [60] R. Fielding, J. Gettys, J. Mogul, H. Frystyk, L. Masinter, P. Leach, and T. Berners-Lee, “RFC 2616: Hypertext Transfer Protocol - HTTP 1.1,” *ISOC Request for Comments (RFC)*, 1999.
- [61] RSnake and John Kinsella, “Slowloris HTTP DoS,” <http://hackers.org/slowloris/>.
- [62] Cloudflare, “What is a Slowloris DDoS attack?” <https://www.cloudflare.com/learning/ddos/ddos-attack-tools/slowloris/>.
- [63] —, “What is a R.U.D.Y. attack?” <https://www.cloudflare.com/learning/ddos/ddos-attack-tools/r-u-dead-yet-rudy/>.
- [64] Akamai, “CDN Definition,” <https://www.akamai.com/us/en/cdn/what-is-a-cdn.jsp>.
- [65] NGINX, “What Is Load Balancing?” <https://www.nginx.com/resources/glossary/load-balancing/>.
- [66] AWS, “What is a DDoS Attack?” <https://aws.amazon.com/shield/ddos-attack-protection/>.
- [67] Cloudflare, “What is DDoS mitigation?” <https://www.cloudflare.com/learning/ddos/ddos-mitigation/>.
- [68] D. J. Bernstein, “SYN Cookies,” <https://cr.yp.to/syncookies.html>.
- [69] T. W. Doepfner, P. N. Klein, and A. Koyfman, “Using Router Stamping to Identify the Source of IP Packets,” *ACM Conference on Computer and Communications Security (CCS)*, 2000.
- [70] X. Liu, X. Yang, and Y. Xia, “NetFence: Preventing Internet

- Denial of Service from Inside Out,” *ACM SIGCOMM Computer Communication Review (CCR)*, 2010.
- [71] J. Ioannidis and S. M. Bellovin, “Implementing Pushback: Router-based Defense against DDoS Attacks,” *Network and Distributed System Security Symposium (NDSS)*, 2002.
- [72] “Cisco,” <https://www.cisco.com/>.
- [73] “Juniper,” <https://www.juniper.net/us/en.html>.
- [74] “Linux Foundation,” <https://www.linuxfoundation.org/>.
- [75] “FreeBSD Project,” <https://www.freebsd.org/>.
- [76] S. Lance, “Honeypots: Tracking Hackers,” *Addison-Wesley Professional - ISBN: 978-0321108951*, 2002.
- [77] B. McCarty, “Botnets: Big and Bigger,” *IEEE Symposium on Security and Privacy (S&P)*, 2003.
- [78] J. Krupp, M. Backes, and C. Rossow, “Identifying the Scan and Attack Infrastructures Behind Amplification DDoS Attacks,” *ACM Conference on Computer and Communications Security (CCS)*, 2016.
- [79] K. Salah, J. M. A. Calero, S. Zeadally, S. Al-Mulla, and M. Alzaabi, “Using Cloud Computing to Implement a Security Overlay Network,” *IEEE Symposium on Security and Privacy (S&P)*, 2013.
- [80] S. K. Fayaz, Y. Tobioka, V. Sekar, and M. Bailey, “Bohatei: Flexible and Elastic DDoS Defense,” *24th USENIX Security Symposium (USENIX Security)*, 2015.
- [81] M. Zhang, G. Li, S. Wang, C. Liu, A. Chen, H. Hu, G. Gu, Q. Li, M. Xu, and J. Wu, “Poseidon: Mitigating Volumetric DDoS Attacks with Programmable Switches,” *Network and Distributed System Security Symposium (NDSS)*, 2020.
- [82] Z. Liu, H. Namkung, G. Nikolaidis, J. Lee, C. Kim, X. Jin, V. Braverman, M. Yu, and V. Sekar, “Jaquen: A High-Performance Switch-Native Approach for Detecting and Mitigating Volumetric DDoS Attacks with Programmable Switches,” *30th USENIX Security Symposium (USENIX Security)*, 2021.
- [83] S. Narayana, M. Tahmasbi, J. Rexford, and D. Walker, “Compiling Path Queries,” *13th USENIX Symposium on Networked Systems Design and Implementation (NSDI)*, 2016.
- [84] M. Elsabagh, D. Fleck, A. Stavrou, M. Kaplan, and T. Bowen, “Practical and Accurate Runtime Application Protection Against DoS Attacks,” *International Symposium on Research in Attacks, Intrusions and Defenses (RAID)*, 2017.
- [85] H. M. Demoulin, I. Pedisich, N. Vasilakis, V. Liu, B. T. Loo, and L. T. X. Phan, “Detecting Asymmetric Application-layer Denial-of-Service Attacks In-Flight with FINELAME,” *USENIX Annual Technical Conference (USENIX ATC)*, 2019.
- [86] Y. Gilad, A. Herzberg, M. Sudkovitch, and M. Goberman, “CDN-on-Demand: An Affordable DDoS Defense via Untrusted Clouds,” *Network and Distributed System Security Symposium (NDSS)*, 2016.
- [87] S. Ramanathan, J. Mirkovic, M. Yu, and Y. Zhang, “SENSS Against Volumetric DDoS Attacks,” *34th Annual Computer Security Applications Conference (ACSAC)*, 2018.
- [88] S. Zheng and X. Yang, “DynaShield: Reducing the Cost of DDoS Defense using Cloud Services,” *11th USENIX Workshop on Hot Topics in Cloud Computing (HotCloud)*, 2019.
- [89] T. Vissers, T. V. Goethem, W. Joosen, and N. Nikiforakis, “Maneuvering Around Clouds: Bypassing Cloud-based Security Providers,” *ACM Conference on Computer and Communications Security (CCS)*, 2015.
- [90] J. Bushart and C. Rossow, “DNS Unchained: Amplified Application-Layer DoS Attacks Against DNS Authoritatives,” *International Symposium on Research in Attacks, Intrusions and Defenses (RAID)*, 2018.
- [91] R. Jansen, T. Vaidya, and M. Sherr, “Point Break: A Study of Bandwidth Denial-of-Service Attacks against Tor,” *28th USENIX Security Symposium (USENIX Security)*, 2019.
- [92] D. Kopp, M. Wichthuber, I. Poese, J. Santanna, O. Hohlfeld, “DDoS Hide & Seek: On the Effectiveness of a Booter Services Takedown,” *ACM Internet Measurement Conference (IMC)*, 2019.
- [93] “Tor,” <https://www.torproject.org/>.
- [94] G. Gu, D. Ott, V. Sekar, and K. Sun, “Programmable System Security in a Software-Defined World – Research Challenges and Opportunities,” *NSF Workshop on Programmable System Security in a Software Defined World*, 2018.
- [95] M. C. Dacier, S. Dietrich, F. Kargl, and H. König, “Network Attack Detection and Defense – Security Challenges and Opportunities of Software-Defined Networking,” *Dagstuhl Seminar 16361*, 2016.
- [96] M. C. Dacier, H. König, R. Cwalinski, F. Kargl, and S. Dietrich, “Security Challenges and Opportunities of Software-Defined Networking,” *IEEE Symposium on Security and Privacy (S&P)*, 2017.
- [97] M. Zhang, G. Li, L. Xu, J. Bi, G. Gu, and J. Bai, “Control Plane Reflection Attacks in SDNs: New Attacks and Countermeasures,” *International Symposium on Research in Attacks, Intrusions and Defenses (RAID)*, 2018.
- [98] M. Zhang, J. Bi, J. Bai, and G. Li, “FloodShield: Securing the SDN Infrastructure Against Denial-of-Service Attacks,” *17th IEEE International Conference On Trust, Security And Privacy In Computing And Communications/12th IEEE International Conference On Big Data Science And Engineering (TrustCom/BigDataSE)*, 2018.
- [99] H. Wang, L. Xu, and G. Gu, “FloodGuard: A DoS Attack Prevention Extension in Software-Defined Networks,” *45th Annual IEEE/IFIP International Conference on Dependable Systems and Networks (DSN)*, 2015.
- [100] G. Shang, P. Zhe, X. Bin, H. Aiqun, and R. Kui, “FloodDefender: Protecting Data and Control Plane Resources under SDN-aimed DoS Attacks,” *IEEE International Conference on Computer Communications (INFOCOM)*, 2017.
- [101] G. Shang, P. Zhe, X. Bin, H. Aiqun, S. Yubo, and R. Kui, “Detection and Mitigation of DoS Attacks in Software Defined Networks,” *IEEE/ACM Transactions on Networking (TON)*, 2020.
- [102] T. E. Carroll, M. Crouse, E. W. Fulp, and K. S. Berenhaut, “Analysis of Network Address Shuffling as a Moving Target Defense,” *IEEE International Conference on Communications (ICC)*, 2014.
- [103] H. Alavizadeh, J. Jang-Jaccard, and D. S. Kim, “Evaluation for Combination of Shuffle and Diversity on Moving Target Defense Strategy for Cloud Computing,” *IEEE International Conference On Trust, Security And Privacy In Computing And Communications/IEEE International Conference On Big Data Science and Engineering (TrustCom/BigDataSE)*, 2018.
- [104] H. Alavizadeh, J. B. Hong, J. Jang-Jaccard, and D. S. Kim, “Comprehensive Security Assessment of Combined MTD Techniques for the Cloud,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2018.
- [105] H. Wang, F. Li, and S. Chen, “Towards Cost-Effective Moving Target Defense Against DDoS and Covert Channel Attacks,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2016.
- [106] A. Clark, K. Sun, and R. Poovendran, “Effectiveness of IP Address Randomization in Decoy-Based Moving Target Defense,” *IEEE 52nd Conference on Decision and Control (CDC)*, 2013.
- [107] A. Clark, K. Sun, L. Bushnell, and R. Poovendran, “A Game-Theoretic Approach to IP Address Randomization in Decoy-Based Cyber Defense,” *Springer International Conference on Decision and Game Theory for Security (GameSec)*, 2015.
- [108] F. Nizzi, T. Pecorella, F. Esposito, L. Pierucci, and R. Fantacci, “IoT Security via Address Shuffling: The Easy Way,” *IEEE Internet of Things Journal (IoT-J)*, 2019.
- [109] S. Yao, Z. Li, J. Guan, and Y. Liu, “Stochastic Cost Minimization Mechanism Based on Identifier Network for IoT Security,” *IEEE Internet of Things Journal (IoT-J)*, 2020.
- [110] P. Kampanakis, H. Perros, and T. Beyene, “SDN-based Solu-

- tions for Moving Target Defense Network Protection,” *IEEE International Symposium on a World of Wireless, Mobile and Multimedia Networks (WoWMoM)*, 2014.
- [111] J. Steinberger, B. Kuhnert, C. Dietz, L. Ball, A. Sperotto, H. Baier, A. Pras, and G. Dreo, “DDoS Defense using MTD and SDN,” *IEEE/IFIP Network Operations and Management Symposium (NOMS)*, 2018.
- [112] Y. Zhou, G. Cheng, S. Jiang, Y. Hu, Y. Zhao, and Z. Chen, “A Cost-effective Shuffling Method against DDoS Attacks using Moving Target Defense,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2019.
- [113] J. H. Jafarian, E. Al-Shaer, and Q. Duan, “Openflow Random Host Mutation: Transparent Moving Target Defense Using Software Defined Networking,” *ACM SIGCOMM - Workshop on Hot Topics in Software Defined Networks (HotSDN)*, 2012.
- [114] —, “An Effective Address Mutation Approach for Disrupting Reconnaissance Attacks,” *IEEE Transactions on Information Forensics and Security (TIFS)*, 2015.
- [115] A. Chowdhary, A. Alshamrani, D. Huang, and H. Liang, “MTD Analysis and evaluation framework in Software Defined Network (MASON),” *ACM International Workshop on Security in Software Defined Networks & Network Function Virtualization (SDN-NFV Sec)*, 2018.
- [116] A. Aydeger, M. H. Manshaei, M. A. Rahman, and K. Akkaya, “Strategic Defense Against Stealthy Link Flooding Attacks: A Signaling Game Approach,” *IEEE Transactions on Network Science and Engineering*, 2021.
- [117] M. Nguyen, A. Pal, and S. Debroy, “Whack-a-mole: Software-Defined Networking Driven Multilevel DDoS Defense for Cloud Environments,” *IEEE 43rd Conference on Local Computer Networks (LCN)*, 2018.
- [118] J. B. Hong and D. S. Kim, “Towards Scalable Security Analysis Using Multi-layered Security Models,” *Elsevier Journal of Network and Computer Applications*, 2016.
- [119] “ONOS,” <https://opennetworking.org/onos/>.
- [120] J. H. Jafarian, E. Al-Shaer, and Q. Duan, “Formal Approach for Route Agility against Persistent Attackers,” *Springer European Symposium on Research in Computer Security*, 2013.
- [121] C. Xu, T. Zhang, X. Kuang, Z. Zhou, and S. Yu, “Context-aware Adaptive Route Mutation Scheme: A Reinforcement Learning Approach,” *IEEE Internet of Things Journal (IoT-J)*, 2021.
- [122] Q. Jia, K. Sun, and A. Stavrou, “MOTAG: Moving Target Defense against Internet Denial of Service Attacks,” *IEEE 22nd International Conference on Computer Communications and Networks (ICCCN)*, 2013.
- [123] H. Wang, Q. Jia, D. Fleck, W. Powell, F. Li, and A. Stavrou, “A Moving Target DDoS Defense Mechanism,” *Elsevier Computer Communications*, 2014.
- [124] P. Wood, C. Gutierrez, and S. Bagchi, “Denial of Service Elusion (DoSE): Keeping Clients Connected for Less,” *IEEE 34th Symposium on Reliable Distributed Systems (SRDS)*, 2015.
- [125] D. Fleck, A. Stavrou, G. Kesidis, N. Nasiriani, Y. Shan, and T. Konstantopoulos, “Moving-target Defense against Botnet Reconnaissance and an Adversarial Coupon-Collection Model,” *IEEE Conference on Dependable and Secure Computing (DSC)*, 2018.
- [126] G. Kesidis, Y. Shan, D. Fleck, A. Stavrou, and T. Konstantopoulos, “An Adversarial Coupon-Collector Model of Asynchronous Moving-Target Defense against Botnet Reconnaissance,” *IEEE 13th International Conference on Malicious and Unwanted Software (MALWARE)*, 2018.
- [127] N. Bandi, H. Tajbakhsh, and M. Analoui, “FastMove: Fast IP Switching Moving Target Defense to Mitigate DDOS Attacks,” *IEEE Conference on Dependable and Secure Computing (DSC)*, 2021.
- [128] S. Venkatesan, M. Albanese, K. Amin, S. Jajodia, and M. Wright, “A Moving Target Defense Approach to Mitigate DDoS Attacks against Proxy-based Architectures,” *IEEE Conference on Communications and Network Security (CNS)*, 2016.
- [129] M. Wright, S. Venkatesan, M. Albanese, and M. P. Wellman, “Moving Target Defense against DDoS Attacks: An Empirical Game-Theoretic Analysis,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2016.
- [130] A. Aydeger, N. Saputro, and K. Akkaya, “Utilizing NFV for Effective Moving Target Defense Against Link Flooding Reconnaissance Attacks,” *IEEE Military Communications Conference (MILCOM)*, 2018.
- [131] S. Debroy, P. Calyam, M. Nguyen, A. Stage, and V. Georgiev, “Frequency-minimal moving target defense using software-defined networking,” *IEEE International Conference on Computing, Networking and Communications (ICNC)*, 2016.
- [132] S. Debroy, P. Calyam, M. Nguyen, R. L. Neupane, B. Mukherjee, A. K. Eeralla, and K. Salah, “Frequency-Minimal Utility-Maximal Moving Target Defense Against DDoS in SDN-Based Systems,” *IEEE Transactions on Network and Service Management (TNSM)*, 2020.
- [133] M. Nguyen, P. Samanta, and S. Debroy, “Analyzing Moving Target Defense for Resilient Campus Private Cloud,” *IEEE 11th International Conference on Cloud Computing (CLOUD)*, 2018.
- [134] Q. Jia, H. Wang, D. Fleck, F. Li, A. Stavrou, and W. Powell, “Catch Me If You Can: A Cloud-enabled DDoS Defense,” *IEEE/IFIP International Conference on Dependable Systems and Networks (DNS)*, 2014.
- [135] W. Peng, F. Li, C. T. Huang, and X. Zou, “A Moving-target Defense Strategy for Cloud-based Services with Heterogeneous and Dynamic Attack Surfaces,” *IEEE International Conference on Communications (ICC)*, 2014.
- [136] S. Venkatesan, M. Albanese, G. Cybenko, and S. Jajodia, “A Moving Target Defense Approach to Disrupting Stealthy Botnets,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2016.
- [137] “CVSS Specification Document,” <https://www.first.org/cvss/specificationdocument>.
- [138] “CVE,” <https://cve.mitre.org/>.
- [139] F. Gillani, E. Al-Shaer, and Q. Duan, “In-design Resilient SDN Control Plane and Elastic Forwarding Against Aggressive DDoS Attacks,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Moving Target Defense (MTD)*, 2018.
- [140] “MATLAB,” <https://www.mathworks.com/products/matlab.html>.
- [141] “Mininet,” <http://mininet.org/>.
- [142] B. Lantz, B. Heller, and N. McKeown, “A Network in a Laptop: Rapid Prototyping for Software-Defined Networks,” *ACM SIGCOMM - Workshop on Hot Topics in Networks (HotNets)*, 2010.
- [143] “Floodlight,” <https://floodlight.atlassian.net/>.
- [144] “DeterLab,” <https://www.isi.deterlab.net/>.
- [145] “Chameleon,” <https://www.chamelecloud.org/>.
- [146] “PlanetLab,” <https://planetlab.cs.princeton.edu/>.
- [147] R. S. S. Kumar, A. Wicker, and M. Swann, “Practical Machine Learning for Cloud Intrusion Detection: Challenges and the Way Forward,” *ACM Conference on Computer and Communications Security (CCS) - Workshop on Artificial Intelligence and Security (AISeC)*, 2017.
- [148] “CAIDA,” <https://www.caida.org>.
- [149] “IMPACT,” <https://www.impactcybertrust.org>.
- [150] “Kaggle,” <https://www.kaggle.com>.