

Instinctive Calibrate based Container System Along with Protection and Database Optimization for Emphatic Cloud based software Testing

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Abstract— Innovative developments of cloud-based application the researchers must conduct cloud-based software tests to assess the reliability and completeness in order to ensure the high quality. Nonetheless, several scholars came up with research on testing technology applied to the cloud, in that there is no specific approach to follow for resource management, software integrity and database configure optimization in order to perform an effectual cloud-based software testing. Hence, the paper proposed a novel Emphatic Cloud Integration Testing with DBM's Framework to support integration of remotely-hosted cloud testing tools in a strong secure and lossless data manner. To begin with reduction of waste resources, the frame work introduces Instinctive Calibrating based Container's system, which performs the implementation of four level mechanism with instinctive calibrate service on containerized orchestration platform to control the calibrate-in/ calibrate-out of containers during work load fluctuation. Along with this for container security and integrity, Isolated Ratification with protection scrutinize Strategy is incorporates that conquer via separate validation to each compute node equipped with a single trusted platform module, and it enables integrity verification of both the host and running containers. At last due to the diverse database instances and query workloads, the framework commences with Tetrad Deep Method to optimize the configurations of database through end-to-end isolated database alteration with attempt-defect manner that overcome the shortcoming caused by regression, hence the proposed work highly reduced the time and space complexity at the occasion of major services as cloud-based software testing.

Key words: Cloud based software Testing, Migration from legacy testing to testing in cloud, Emphatic Cloud Integration Testing (ECIT), Instinctive Calibrating based Containers system (ICCS), Isolated Ratification with protection scrutinize Strategy (IRPSS), Tetrad Deep Method (TDM).

I. INTRODUCTION

Cloud computing systems-based Computational architectures are probably the most cost-effective approach for [1] many end-users: businesses and scientists. However, when delivering a device with features such as 24/7 availability, worldwide use, and convenient access for any user, a variety of factors have to be handled. Although the anticipated [2] functionality can be provided by a cloud system, issues such as efficiency and resource provisioning are also essential, as a data center [3] would include a large number of computers and

communication networks. As cloud services are built to be scalable, the significance [4] of this problem is only likely to rise, with the possibility of adding resources [5] to increase the total system ability. In particular, in cloud computing environments, the issue of resource provisioning is complicated by the need for end-users not to note a performance loss.

When designing cloud computing services, one of the key challenges that we have to solve is to ensure that the system's [6] behavior is compatible with standards. Here, the system's

action requires variables like efficiency and user management. Checking [7] is probably the most commonly used tool for validating the correctness of programmers. If from a formal model we start the development of a method, then testing can be used to conduct more rigorous analysis [8]. It is, however, a major challenge to establish systematic testing methodologies for cloud systems [9].

Normally, a cloud system would include several interacting and distributed components, making it difficult to implement systematic testing [10] approaches. It is common to use an oracle in testing to verify that the behavior observed during testing is permissible / acceptable in a given test case. An oracle can be a realization [11] of a formal system specification or a set of properties that must be satisfied by the system. In certain cases, however, an oracle is not usable or is too [12] costly to implement in terms of computation, and alternate methods must be used [13]. This is especially troublesome when testing complex systems, where whether authors are supposed to fully verify the system's actions, several test cases [14] should be produced and executed. In cloud computing, in which there is rarely an [15] oracle and massive test suites are needed to verify the crucial elements of the program, certain difficulties arose.

In addition cloud testing faces cost and scalability challenges due to the increasing extent and complexity of the software. However, testing tools [16] can also take advantage of a cloud infrastructure's massive resources, like virtualized systems and facilities, broad processing power and memory, concurrent operations, and automated recovery mechanisms [17] that come in a PaaS. This research accomplished a simulation based framework, the key objective is to provide a tool-supported methodology that enables users to design both the software and hardware components of cloud systems, implement new cloud systems, and evaluate these systems automatically using a cost-effective approach that takes into account both functional and non - operational attributes of the cloud. Therefore, the development of successful systems to achieve automated configuration and optimization of parameters in integration testing tool becomes an indispensable way to address major obstacle. Consequently, from above mentioned things the paper develop a proficient system which overwhelmed the significant concerns in emerge field.

II. LITERATURE SURVEY

Morán et al [18] In Big Data Engineering, modern processing models is being implemented to solve the limitations of conventional technology. Map Reduce stands out amongst them by enabling vast amounts of data to be processed over a distributed infrastructure that can alter during runtime. The developer only designs the program features and a distributed

framework handles its execution. As a result, at each execution, a program will behave differently because it is adapted automatically to the resources available at each moment. Hence, this may be exposed in some executions and disguised in others when the program has a design flaw. However, these faults are typically masked during testing because the test system is reliable and they are only exposed in production because, among other factors, the environment is more hostile due to infrastructure failures. This paper suggests new research methods aimed at stimulating various infrastructure configurations to detect these architecture faults. Testing techniques produce a representative set of infrastructure configurations that, along with combinatorial testing, are more likely to expose deficiencies using random testing and partition testing.

Hieronset al [19] the device under test communicates with its environment at several physically dispersed ports in the dispersed test architecture and the local testers at these ports do not synchronize their behavior. In particular, seemingly incorrect acts may be the result of a mistaken belief about the exact order in which actions were conducted at various ports. This poses several challenges. The paper defined a conformance relation for the distributed test architecture in previous work. Essentially, if they observe a trace σ such that no admissible reordering of the acts in σ could have been generated by the specification, the device under test is defective. This notion, however, can be weak if the traces compared might be too different. This paper addresses conformance relationships where a reordering is considered for a given metric only if the distance between the two traces is at most a certain bound k . the research introduce two different metrics and provide algorithms for the construction of finite automata that accept these near sequences with regard to each metric. The computational complexity of the two key problems associated with the new architecture is also studied to determine whether one device complies with a specification with regard to the new conformance relationship. Jin et al [20] the combination of container technology and micro service architecture makes the container-based cloud environment more efficient and agile than the cloud environment based on VM, due to the lightweight features. But it also greatly amplifies the dynamism and complexity of the cloud environment and simultaneously increases the uncertainty of system security issues. In this case, as the updates take place in cloud environment, the effectiveness of defense mechanisms with fixed strategies would fluctuate. They refer to this problem as effectiveness drifting issue of defense mechanisms, which is particularly acute in proactive defense mechanisms, such as moving target defense (MTD). To tackle this issue, they present DSEOM, a framework that can automatically perceive container-based cloud environment

updates, quickly assess MTD's change in effectiveness and dynamically optimize MTD strategies. Specifically, developers are creating a multidimensional model of attack graphs to formalize various complex scenarios of attack. In combination with this model, researchers introduce the concept of between's centrality in order to effectively assess and optimize MTD's implementation strategies. Moreover, authors present a series of safety and performance metrics to quantify the effectiveness of DSEOM's MTD strategies.

Yao et al [21] with the rapid growth of information technology, there is a marked increase in cloud computing research activities. Cloud testing can be interpreted as I testing cloud applications involving ongoing monitoring of cloud application status to verify service level agreements, and (ii) testing as a cloud service involving the use of cloud as a testing middleware to perform a large-scale simulation of real-time user interactions. This study is aimed at examining the methodologies and tools used in cloud testing and the current trends in cloud computing testing research.

Nurul et al[22] Because the program bugs were discovered and patched in STLC the earliest, the smaller the sum required to repair them. With the advent of cloud computing, a lot of new business opportunities are opening up, particularly in the field of software testing & maintenance. System is to be followed while cloud service testing is underway. A new approach to test the cloud, known as the SUPerB approach, discusses cloud protection, user adoption, efficiency, and business criteria is introduced. Successful implementation of SUPerB cloud testing methodology could lay a strong foundation for a market leading organization. Vulnerabilities in software give rise to the cyber-crime and related risk associated with it simply due to lapses in security policies, which increases the company security breaches. In terms of performance testing the efficiency of a system can be measured.

Ebadi et al [23] Cloud computing are a kind of parallel, configurable, and scalable network that refers to the provision of virtual data center applications. However, it has become timely and important challenges to reduce the energy consumption and also to maintain high computation capacity. To face those challenges, the concept of replication is used. Increasing the number of data replicas also raises the energy usage, the efficiency and also the cost of produce the problem is formulated in this paper as an optimization problem, and it is offered a hybrid met heuristic algorithm to solve it. The algorithm uses the global search functionality of the Particle Swarm Optimization (PSO) algorithm, and the Tabu Search (TS) local search functionality to obtain high-quality solutions. The method's efficiency is demonstrated by contrasting it with the basic algorithm PSO, TS, and Ant Colony Optimization (ACO) on various test cases. The findings obtained show that

they are both outperformed by the system in terms of energy and expense consumed. Deciding on the number and location of required replicas on the cloud system is an NP-hard issue. In [18] attain the degradation of containers performance [19] offered over supplying therefore there is chance of lack of resources [20] poor security monitoring [21] does not maintains the software integrity [22] get a possibility of vulnerabilities during host runtime [23] much more time complexity [24] does not optimized the database configuration. Thus, from the aforementioned issues, the research work to develop a novel framework in the emerge field of cloud testing.

III. EMPHATIC CLOUD INTEGRATION TESTING WITH DBM'S FRAMEWORK

Testing in cloud is an on-demand platform of Cloud Services and Providers (CSPs), thus its alternatives the traditional system of using local servers or personal computers with a network of remote servers hosted on the internet. Testing tools in cloud environment can be performed at three levels of granularity: Unit Testing is validating the functionality of a component in isolation; Integration Testing is validating the interactions among multiple components; and System Testing is validating the entire piece of software to ensure functional and non-functional requirements are being met. The overall architecture of cloud testing environment is described in figure 1.

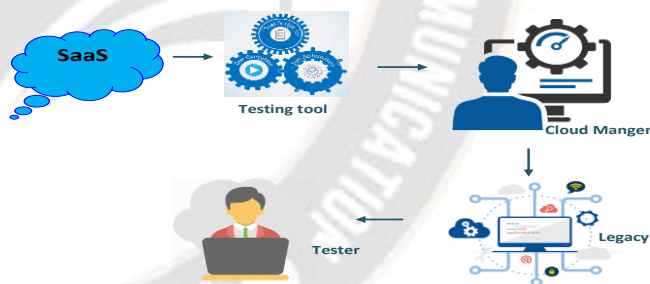


Figure: 1- over all architecture of cloud testing environment

Mostly on other side, large enterprises such as IBM, Microsoft, Google, and Amazon have used important cloud computing services. Researchers, on either side, are continually evolving tools and techniques to strengthen this new technology. Several factors make reasoning especially difficult about a cloud system's underlying architecture. First of all, cloud systems are very broad and this hampers these systems' evaluation. Second, the cloud services given to end-users are virtual and this complicates the research since it is possible to host multiple VMs on a single computer, sharing a resource between various users. Users can build a cloud model and a workload to be tested. The configuration of the

components defining a cloud system is, basically, a cloud model. Among other things, this design involves network topology, physical machine aggregation in racks, physical machine description, VMs, and software components such as managers and schedulers. A workload depicts the tasks that the cloud must process. These operations consist, in turn, of deploying the user-requested VMs and performing requests over the VMs. In that permeability in container is the fundamental building block of testing that is a critical feature of cloud platform, which efficiently performs the reduction of runtime costs. Hence, a model of permeability to be interactions with supply and demand that computes resources in the cloud environment. The former systems which utilized the permeability aim to develop prediction models, in order to allocate resource in advance. Also, those mathematical models first should be trained by real workloads. While permeability contributes that significant to measure the performance of lightweight containers, over-supplying causes the waste of resources. In addition a lot of attention is currently being received to Lightweight containers that have been a pervasive approach to help fast develop, test and update the cloud/IoT. Existing container technologies works well with physical platforms, but not so well with virtual machines hosted in a full virtualization environment (such as the Xen hypervisor or Kernel-based Virtual Machine) and it is simply not available for a lightweight virtualization environment (such as container). Therefore, existing solutions to ensure the integrity of data, but they do not cover the whole service lifetime, as the data and its internals may be changed at run-time by the host or by external attackers exploiting some vulnerability. For instance, an attacker that has gained a privileged access to a running container may compromise it by modifying service configurations and binaries, launching malicious scripts, or starting new processes, all without being detected by static verification techniques. After protection of containers, optimizing is essential for performance of storage configurations. It becomes more tedious and urgent for cloud storage due to the diverse database instances and query workloads, which make the database administrator (DBA) incompetent. Although some studies of automatic configuration of storage exist, they have several limitations on tuning. For example, they follow a pipeline learning model but cannot optimize end-to - end overall efficiency. Then they rely on high-quality, large-scale training samples which are difficult to obtain. Ultimately, there are a huge number of knobs in continuous space and with unknown dependencies, so they cannot suggest rational configurations in such a high-dimensional continuous space. This research mainly focuses on enabling automated integration testing tools in terms of time and space complexity in cloud platform and services.

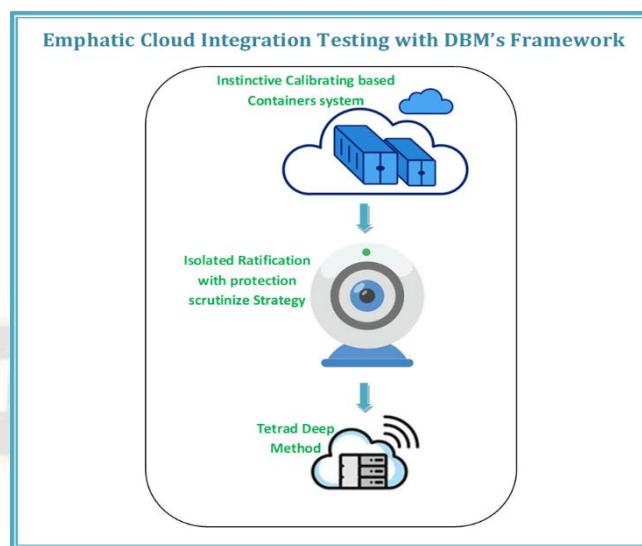


Figure 2- Block Diagram of novel Emphatic Cloud Integration Testing with DBM's Framework

Cloud testing tool has gained considerable attention recently and has begun to attract corporations, general users and developers. Nonetheless, several scholars came up with research on testing engine technology applied to the cloud based software testing, in that there is no specific approach to follow for resource management, software integrity and configure optimization in order to perform ineffective testing. In order to rapidly test the cloud, the test obtain consent as input a cloud model, a workload, and a collection of MRs, previously chosen by the user. This framework basically carries out the testing process. The following procedures are then executed routinely. Initially, a list of workloads is developed. Such workloads are created automatically by the user initially provided from the workload. The testing engine proceeds to produce the tests once the workloads are generated. First, a test suite is created automatically by using both the user-generated cloud model and the workload set. A test case consists of a cloud model and a workload that the cloud can process. Second, by slightly changing the cloud model supplied by the customer, a wide number of variants (clouds) are generated. Next, by combining the variants (clouds) and the workloads, a wide set of follow-up test cases is automatically generated. The input portion of the MRs involved in the testing process must be satisfied by each test case generated and each follow-up test case generated. Third, all research cases are carried out on a model for simulation. In this paper present a novel Emphatic Cloud Integration Testing with DBM's Framework to support testing of applications tools that depend on remotely-hosted cloud services that perform through proficiently tackles the prior significant issues. Initially to measure the performance of

lightweight containers with reduction of waste resources, the frame work introduce Instinctive Calibrating based Containers system, which performs the implementation of instinctive calibrate service on containerized orchestration platform, thus it consisting of four level of mechanism such as monitoring mechanism, history recorder, decision mechanism and execution mechanism, to control the calibrate-in/ calibrate-out of containers during work load fluctuation. Thus, the novel system enhanced the automatic resource management with less time complexity. Along with this, container security and integrity are huge vital after resource management, in prior techniques does not cover the lifetime of whole cloud services as external attackers exploiting some vulnerability during run time. To overcome this, the frame work establishes a novel Isolated Ratification with protection scrutinize Strategy that can be applied separately to each compute node equipped with only a single trusted platform module like it is a signature-based strategy, and it enables integrity verification of both the host, container engine and running containers. The evidence of the integrity state of the services running inside these containers is authenticated by the module, which fulfills the requirement of software integrity permission in lightweight cloud environment. At last due to the diverse database instances and query workloads, the framework commences with Tetrad Deep Method to optimize the configurations of database in high dimensional continuous space. Thus, it accomplishes through end-to-end isolated database alteration with attempt-defect manner that overcome the shortcoming caused by regression, consequently optimized the database alteration in continuous space, hence it highly reduced the time and space complexity at the occasion of major services as database testing. Accordingly, from the proposed novel framework expertly overcomes the drawbacks in prior cloud integration testing tool in an effectual manner. The detail description about the proposed framework is follows.

3.1. Instinctive Calibrating based Containers system:

Two aspects of the information are considered by the container-based instinctive calibrator: the requests obtained and the accessible computing resource revealed by Resource Management, e.g. Memory and Processor. The instinctive calibrator allocates the corresponding compute resource to the provision or de-provision of the containers across Cloud Computing nodes, taking everything into account. In short, as a standalone and third-party service, the instinctive calibrating section controls the usage of each container's real-time computing resource and adopts user-defined calibrating policies, makes calibrating decisions and performs calibrating behavior.

The strategy consists of five parameter values considered by baseline calibrating. First of all, the lower and upper computing resource thresholds are significant, in the form of a percentage. The instinctive calibrator starts to record the timestamp when the thresholds are exceeded. Unless the resource usage has been held above the limits for t^S , the instinctive calibrating operation will be carried out. If after the last instinctive calibrating operation the over-provision or under-provision persists, it takes time to wait t^U and then automatically scale again. The size of VMs and host PM is correlated with the largest number of containers, so make it dependent on the realistic situation. The container range in the experiment is [0, 5]. Online measurements will be continuously documented and decisions will be taken by them. The configuration of the instinctive calibrator and the theory are shown in figure 3. It consists of four elements, respectively, the monitoring system, history recorder, and decision system and execution mechanism. In the following sections, they are elaborated further in depth.

3.1.1. Monitoring system:

The container-based instinctive calibrator first sends the pulse message to the DC/OS scheduler and retrieves all the applications' data. It tests its tag field for each application and figures out whether a calibrating policy is established. If so, it will parse the concept of the calibrating policy and start polling the corresponding metrics on the master node.

```

Algorithm 1 Monitoring system
Input : D(r), Y(r), K(r), M
-----
If  $(D(r) < K(r)) \parallel (D(r) < Y(r))$ 
    Transfer the information to record the history.
Else
    stay behind unmovable
End if
Transfer the suitable time concert report
    
```

A list of executors/ containers under this same program, with the use of the CPU and memory, is retrieved after doing this. It waits for a short period of time $T\Delta$ for the second poll after the first poll, once monitoring then performs to evidence the history.

3.1.2. Record history:

In this section, management of the calibrating history and its start / finish timestamp points is taken into account. At any

point, if the paper finds that an application has a metric running above its upper calibrating threshold, this event is reported in a history recorder along with the timestamp. Users are permitted to specify the time of in-duration. If throughout the in-duration time there are sufficiently many such above-calibrating-threshold events (t^S), it indicates that user concern is highly exploited by the compute resource. The container instinctive calibrator will send out an output warning signal immediately at this point and will therefore consider a calibrating operation, at time t^S , the first calibrating begins and ends at time ($t^S + T$). It is during this cycle of calibrating time, however, that the application demonstrates its worst performance stability. The paper therefore considers a relatively long period of time, t^U . That makes it possible to cool the application down before contemplating the next cycle of the calibrating phase.

Algorithm 2 Record History

Input: t, t^U, t^S

Output: t_1 - The timestamp recording the primary moment instinctive calibrator obtain alerts, t_2 - The timestamp while calibrating exploit is elicited, t_3 - The timestamp once the settle down period ends.

If there is the first time receiving signals,

$t_1 = t,$

Else if $(t - t_1) \geq t^U$

$t_2 = t$ && transfer information to decision strategy;

Else if $(t - t_2) \geq t^S$

$t_3 = t$ && transfer information to decision strategy;

End if

Users may specify the in-duration period and the cool-down period, but if shorter than the calibrating period T , pay attention to the cool-down period, leading to very unpredictable calibrating behavior that users would lose control of the instinctive calibrator and could not properly comply with the strategies set. The History Recorder method is presented in Algorithm 2. The definitions of t_1, t_2, t_3 are: no calibrating is anticipated during t_1 , since there are not enough counts of warnings that the measurement resource has surpassed the threshold; calibrating action is triggered during t_2 , since all the required conditions are met; no calibrating is triggered during t_3 , since it is still within the cool-down time of the last calibrating time.

3.1.3. Decision strategy:

Each user pre-defines the instinctive calibrating approach according to the need for their system configured within the containers. The CPU use of the containers is frequently polled by the instinctive calibrator. As the Decision Process is seen in figure 3, even though the resource consumption metric has surpassed its threshold for some time during the t_1 cycle, there are inadequate counts of such events during the t_2 cycle, which does not contribute to any calibrating intervention. However, the time of t_2 met all three requirements, and a calibrating action is required from M containers to M' containers. The framework is still within its cool-down duration during the time of t_3 ; thus, no scaling action must be carried out. In depth, entire process could be described in Algorithm 3.

Algorithm 3 Decision Strategy

Input: $\{ \langle t^S, t^U \rangle, \langle K(r), I(r) \rangle, \langle m^{\min}, m^{\max} \rangle, \Delta m \}$

Output: M'

Transfers $\langle K(r), I(r) \rangle$ to observing strategy;

Transfers $\langle t^S, t^U \rangle$ to record history;

If recorded histories transfers information's to verdict strategy,

$M' = M \pm \Delta m;$

Else if

$((M' \geq m^{\min}) \&\& (M' \leq m^{\max})),$

modify the integer of containers to M' ;

Transfer the timestamp and action to record the history;

Else

continue unmovable;

End if

3.1.4. Execution strategy:

If the instinctive calibrating container eventually wishes to scale, it would then submit a query to Revolution, indicating the new container which the service wants to use it until the request has been received, and will then send an upgrade request to the master nodes to expand the cluster. To illustrate the procedure, three steps are mentioned in the following subsections.

a) Calibrating out and calibrating in:

Calibrating out a container community requires the need to check whether all slave nodes have enough computing resources to support the new cluster for the leading master node. If the request is not rejected immediately and the calibrating operation fails, the program will initialize a resize deployment otherwise, and new executors / containers will be deployed to the slave nodes. Calibrating in is a little complicated since all the containers are likely to be used to supply workloads. The simplest option will be to have a grace period for the containers to be recycled before they can be executed. By removing the containers from either the service composition framework, it can also be overcome and all queries for such containers are reprocessed to the active containers which will not be destroyed.

b) IP and ports for publish containers:

The newest containers are distributed after such a calibrating-out action. This one would report the IP address and routing port of its hosting system, and then following the diagnostic test of the new containers passed by the system, upcoming queries might be workload adjusted to the fresh containers.

c) Containers maintenance:

Two levels of organizational-control systems are available. The first level begins from the service viewpoint, which develops a wellness-checking endpoint for each service provider and tests if the service is adequately deployed within the container. If this series of tests fails, many approaches to deal with this problem can be enforced by the service provider, such as re-launching the system within the container, etc. At the container level, the second level series of tests occurs. The system will destroy the container and reallocate the container to another host if a container fails to run for a certain period of time. Thus the novel system enhanced the automatic resource management with less time complexity. Along with this, container security and integrity is huge vital after resource management in testing engine of cloud, which is described in upcoming section.

3.2. Isolated Ratification with protection scrutinizes Strategy:

Container frames may be used by an attacker to hold un-trusted software or to embed malware. In addition, software for which a flaw is identified at any point can be used by the file maker. In this case, the insecure software will be used in all the containers instantiated from it unless the file is re-built in the container file provider (e.g. the Docker Specific File

Repository). Then, additional risks can be introduced by the object delivery process. Docker implements a hosting site registry service that can be abused to retrieve images from a centralized entity through multiple container engines operating on different hosts. The risk of hacker-in-the-Middle attacks and compromised applications can be increased by insecure connectivity to the registry, poor server and client authentication, and lack of automated monitoring of stale files. In addition, container orchestration platforms can introduce security vulnerabilities in the event that the administrator does not properly manage them. The orchestrator can, more precisely, ensure that inter-container interface is perfectly designed and that workloads with different levels of sensitivity are not mixed. In addition, the improvisation platform's trustworthiness should be validated. Finally, if the container engine itself (for example, the Docker daemon) is improperly designed or has been abused, vulnerabilities can be introduced. In this respect, along with its runtime configuration, the integrity of the container engine should be checked to ensure that a malicious user does not manipulate the container engine itself to acquire host privileges. Because of this, it is also important to verify the host OS itself, and its attack surface should be limited as much as possible to avoid vulnerable software execution and inappropriate configuration.

The consequences can target the integrity or interaction of each container with other containers or the host OS. In order to contrast these issues, an integrity assurance system should be in place to ensure the trustworthiness of the following entities: the picture of the runtime container that can carry out both out-to-date and un-trusted applications, or even malware; the container engine itself that an intruder might have compromised to exploit a vulnerability; the underlying host OS that can be compromised in case manages the attacker to exploit a container undetected vulnerability to the OS attack. A static and simple approach to integrity verification is natively provided by Docker, as it can only verify the integrity of the file at load time. This is offered by Container Functionality Assurance (CFA), implemented in version 1.8 of Docker, to allow its developer to verify the file system against the digital signature. Then the research accomplished the trusted platform infrastructure module for integrity assurance is elaborate.

3.2.1. Assurance of integrity based on trusted platform infrastructure:

Three essential features must be introduced by a Trustworthy Platform (TP): stable capabilities, assessment of integrity, and reporting of integrity. Protected capabilities are commands with exclusive licenses to function on shielded sites, i.e. special platform regions where confidential data can be

processed and run safely. The shielded positions reside in the TPM's internal memory in the solution proposed by TPI and are called emergency preparedness Variables (EPVs). Each TPM has a minimum of 24 PCRs, numbered from 0 to 23, by default, and each EPV has a size of 20 bytes, that is, the size of the SHA-1 digest. Several versions of the TPM specifications have been developed by the TPI, the latest being version 2.0. This introduces different EPV banks, in which the same hash algorithm is used to expand all registers, i.e. Algorithm 1 (SHA-1) for Safe Hash. Platform integrity is characterized as a collection of metrics that identify components of the platform, such as the OS, applications and their configurations. A fingerprint serves as a unique identifier for each component and as evidence of no change; components of software are "measured" by a binary attestation approach, i.e. by computing the cryptographic digests on the file over their binaries. Later, to resolve the restriction on the total number of protected registers, the calculation mechanism uses protected capabilities to "cumulatively" store these platform metrics in the EPVs. Integrity Calculation is the activity used to determine the confidence level of the device: before passing device control to it, each component in the system start-up phase tests the to-be-loaded component and these tests are stored in the PCRs. This process of measurement is known as transitive confidence and forms the basis for Remote Attestation. The protected capabilities allow users of the platform to free-read access to the PCRs, but direct writing is prevented. Therefore, EPVs function as accumulators: whenever the validity of a register is changed, the new value relies on both the new amendment and the old value, to ensure that the value of an EPV cannot be altered until it has been initialized. This is the method of extension that operates as follows:

$$EPV_{new} = SHA_function(EPV_{old} || measured_information)$$

(1)

Where, EPV_{old} seems to be the value present throughout the register before the extend operation. The SHA function is the symmetric block cipher (i.e. SHA-1 in the case of TPM version 1.2, both SHA-1 and SHA- in the case of TPM version 2.0), is the cryptographic hash function (i.e. SHA-1 in the case of TPM version 1.2, both SHA-1 and SHA- in the case of TPM version 2.0), and the calculated data is the new measure to be added. Most EPVs (typically EPV-0 to EPV-15) have persistent values before resetting (e.g. rebooting) the entire platform. Thus, except in the case of a system compromise, once the system is rebooted, an attacker will not forge EPV values in his favors. At a particular time, the EPV value is the product of structured extension functions conducted on the parameters of the platform, and the cryptographic nature of this operation makes it difficult to recover any specific interface status calculation. As a result, integrity assessment

architectures usually log each integrity measure in order to start from the individual digests to recreate the final EPV value. The TPI identifies three so-called Source of Confidence (SoC) components to trust the operations of protected capabilities, i.e. components of a TP that are supposed to be trustworthy since their misconduct could not be identified. The Source of Confidence for Measurements (SCM) incorporates an engine that is capable of producing intrinsically accurate measurements of honesty, and it is also the root of the transitive trust chain. The Source of Confidence for Storage (SCS) maintains the measurements of integrity (or a description and sequence of those values) safely and preserves the data and cryptographic keys used by the TP kept in external storage. The Source of Confidence for Reporting (SCR) is capable of reporting to external entities (e.g. users of the platform) the measurements kept by the SCS reliably. Both SCS and SCR can be supported by a TPM, while the BIOS usually implement the SCM and it is triggered as soon as the device is booted. Therefore, EPVs function as accumulators: whenever the validity of a register is changed, the new value relies on both the new amendment and the old value, to ensure that the value of an EPV cannot be altered until it has been initialized.

3.2.2. Calculating Authenticity Design (CAD) for Runtime:

A TPI-compliant Integrity Calculation Record is maintained by CAD and it is the state-of-the-art of static calculation frameworks. It is the first realistic work to expand the principles of TPI confidence calculation to dynamic executable information from the BIOS all the level up to the application layer, making the Virtual authorization technique in standardized frameworks often more appropriate. In this sense, a real-world situation makes a TC-compliant device realistic. Upon enabled, CAD will actually measure the obtained files according to the requirements stated in the administrator-defined application policy. It may limit the form of observable events, such as memory mapped files and executable files, using the CAD policy. Before the measured component takes charge of the platform, each digest is determined and its value extended into an EPV, ensuring that the measurement obtained cannot be manipulated by the component itself. CAD retains the integrity calculation aggregation over one of the available registers in the EPV, i.e. by default, EPV-10. The latter will be included in the evidence provided by the EPV, so that it is possible to verify the trustworthiness of the measurement list cryptographically. Using the CAD measurement list along with the anticipated EPV values, a challenging party should attest to device runtime integrity. By hashing the concatenation of each

measurement and the previous hashed value, the verification process simulates the extension procedure. If the result matches the EPV-10 checked value, it can be inferred that the list of CAD measurements has not been tampered with. By design, CAD does not translate easily into a virtualized environment. The information on the integrity state created by the services running within a virtualized instance cannot be directly applied to the TPM hardware, thus raising security concerns. For instance, even if CAD is enabled within the VM, the TPM can not explicitly or implicitly authenticate the CAD measurement list in a scalable way. Problems are either the lack of direct contact between the VM OS and the TPM, or the lack of adequate EPVs to accommodate the amount of VMs that normally run on a server. For these purposes, at the virtualization layer, the chain of trust is broken and the TPM can only be used to testify to the credibility of the hypervisor. This is the strategy of Trustworthy Compute Pools, which gives confidence in the computing nodes, but not in the hosted VMs. In this work, we are focusing on allowing IMA support for containers that use host kernel sharing differently than VMs. Because of this, irrespective of the virtualization layer, the research is able to hold the chain of trust intact. Then subsequently diverse database instances and query workloads, there is a great essential to optimized the configurations of database in high dimensional continuous space.

3.3. Tetrad Deep Method:

The proposed tetrad deep method is based on database testing this is a kind of cloud software validation that tests the configuration, records, controls, etc. of the database being checked. It tests the accuracy and quality of data as well. It can require creating complex queries to test the database for load / stress and verify its responsiveness. In addition, which is a structural type of database testing that authenticates all the entities that are primarily used for data storage within in the data repository but these are not permitted to be manipulated explicitly by end-users. In structured database testing, the verification of application servers is also an influential concern. The novel method is a regulation based method and can straightforwardly discover the regulation. In certain terms, rather than computing and storing the related Q-values for all actions, the tetrad deep approach is capable of instantly acquiring the unique value of the actual continuous action according to the current state. Therefore, through high-dimensional states and behavior, specifically internal measurements and knob configurations, the proposed method will learn how to regulate. In this section, the paper initially incorporate the regulation based concept, then portrayed the remuneration functions.

3.3.1. Deep Regulation ascent:

First, authors considering the case that remains to be optimized as environment W, and at time i, the adjustment agent will obtain normalized internal metrics a_i from W. Then the adjustment agent generates the knob settings b_i and, after deploying at on the case, will receive a remuneration e_i . It has

a parameterized policy function $b_i = \lambda(d_i | \rho^\lambda)$, similar to

most regulatory ascent techniques, (ρ^λ mapping the state d_i to the action value b_i which it is normally called player). The

network's essential function $\Phi(d_i, b_i | \rho^\Phi)$ is (ρ^Φ , learnable parameters) seeks to reflect the value (score) with particular action at and state d_i , which guides actor learning. In particular, the critical function helps determine the settings of the knob created by the actor according to the current state of the case. The inheritance of Bellman Equation insights and the expected $\phi(d, b)$ is defined as:

$$\Phi^\lambda(d, b) = H_{e_i, d_{i+1} \approx H} [e(d_i, b_i) + \phi \Phi^\lambda(d_{i+1}, \lambda(d_{i+1}))] \quad (2)$$

Where, the regulation $\lambda(d)$ is deterministic, (d_{i+1}) is the state of next order, $e_i = e(d_i, b_i)$ is the remuneration function, and ϕ is a reduction factor which indicates the significance of the future remuneration correlate to the current remuneration.

During parameterized by β^ϕ , the representation of critical value is $\Phi^\lambda(d, b | \rho^\Phi)$ under the regulation λ . Then transitions of sampling (d_i, e_i, b_i, d_{i+1}) from the respond memory, the research work apply algorithm and reduce the objective of training:

$$\text{mink}(\rho^\Phi) = H \left[\left(\Phi(d, b | \rho^\Phi) - u \right)^2 \right] \quad (3)$$

Where

$$u = e(d_i, b_i) + \phi \Phi^\lambda(d_{i+1}, \pi(d_{i+1}, \lambda(d_{i+1}))) \rho^\Phi \quad (4)$$

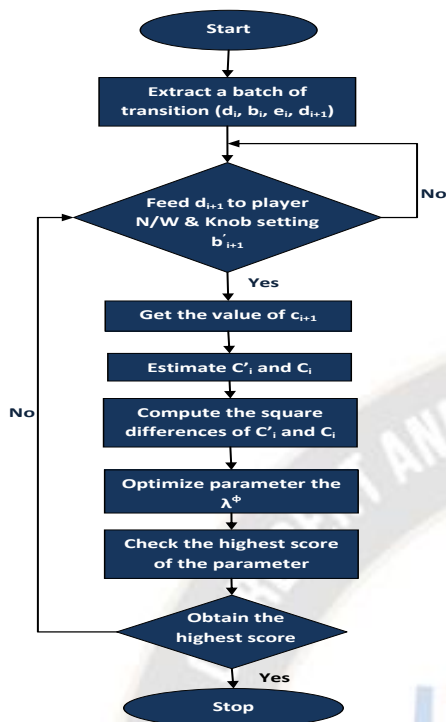


Figure: 3-Flow cart of proposed optimization strategy

Essential parameters can be modified with sigmoid function. As for the player, the chain rule will be applied and modified

with the significant fraction derived from $\Phi^\lambda(d, b | \rho^\Phi)$:

$$\nabla_{\rho^\lambda} T \approx H \left[\nabla_{\rho^\lambda} \Phi(d, b | \rho^\Phi) \Big|_{d=d_i, b=\lambda(d_i)} \right]$$

(5)

$$= H \left[\nabla_{\rho^\lambda} \Phi(d, b | \rho^\Phi) \Big|_{d=d_i, b=\lambda(d_i)} \nabla_{\rho^\lambda} \lambda(d | \rho^\lambda) \Big|_{d=d_i} \right]$$

(6)

The proposed optimization of database configuration in cloud test engine is accomplished based on the procedure of figure 4 flow chart. After optimize the parameters of the database configuration then to perform the remuneration of appropriate function that is explained in below.

3.3.2. Remuneration functions:

In the Tetrad deep system, which provides impactful input information between the agent and the environment, the remuneration function is critical. Authors want remuneration to simulate scientific judgment in the optimization procedure for the modern environment. Optimization process as follows,

- (1) Suggest that T_0 is the original DBMS output and T_i is the final tuned output.
- (2) The proposed work tunes the knobs and, after the first tuning, the output becomes T_1 . The shift in output is then determined (T_1, T_0).

(3) It expects the current output to be better than the previous one at the m -th tuning iteration (i.e., T_m is better than T_{m-1} where $m < I$ since the paper aims to boost output through optimization. It cannot, however, ensure that T_m is better at any iteration than T_{m-1} . The work compares (a) T_m and T_0 and (b) T_m and T_{m-1} to this end. The tuning pattern is right if T_m is better than T_0 , and the remuneration is positive; otherwise, the remuneration is negative. Based on $\Delta(T_m, T_0)$ and $\Delta(T_m, T_{m-1})$, the remuneration value is determined.

Thus it accomplishes through end to end isolated database alteration with attempt-defect manner that overcome the shortcoming caused by regression, consequently optimized the database alteration in continuous space, hence it highly reduced the time and space complexity at the occasion of major services as database testing.

From the aforementioned concerns of the proposed framework in the emerging field of cloud based software testing overwhelmed prior issues such as waste of resources, software integrity and database optimization in testing engine of the cloud through accomplishment of containerization with four level of mechanism, single trusted platform module and integrity verification of both host and running containers, then end to end isolated database alteration with attempt-defect manner in order to that the proposed work proficiently achieve. Then the competence of the novel work proved through experimental evaluation that is performed in upcoming section.

IV. RESULT AND DISCUSSION

Cloud Testing is a concept used to describe experiments that are conducted using cloud infrastructure, i.e. authors do not have to locally install hardware or services and can use the cloud infrastructure on demand for the research testing. The use of cloud testing makes it simple to create an environment for the test in the case of performance testing. Overall, which accomplished the simulating users around the globe reduces time and expense.

4.1. System requirement:

The proposed framework comprehensive study and its performance are analyzed in this section. In the paper, the experiment under taken based on SOASTA Tool that attains parallel test repeat type and test repeat method is fixed with number of repeats are 25 and time interval is 1200ms. The proposed method is implemented in the working platform of MATLAB with the following system specification.

Platform	MATLAB 2017b
OS	Windows 8
Processor	Intel core i7
RAM	8 GB RAM

In the proposed work consists of following test cases, based on that cases the work attains efficient performance in terms of time and space complexity of cloud platform and their services:

- 1) Verify response time is not more than 4 secs when 1000 users access the website simultaneously.
- 2) Verify response time of the Application Under Load is within an acceptable range when the network connectivity is slow
- 3) Check the maximum number of users that the application can handle before it crashes.
- 4) Check database execution time when 500 records are read/written simultaneously.
- 5) Check memory usage of the application and the database server under peak load conditions
- 6) Verify response time of the application under low, normal, moderate and heavy load conditions.
- 7) Verify the network speed is not less than 5 mbps when 1000 users access the website simultaneously.
- 8) Check memory usage of the application and the database server under normal load conditions
- 9) Verify response time of the Application Under Load is within an acceptable range when the network connectivity less than 5 mbps
- 10) Check the maximum number of users that the network connectivity less than 5 mbps.
- 11) Verify response time of the network connectivity less than 5 mbps and heavy load conditions.
- 12) Check the Error Count that the network connectivity less than 5 mbps and heavy load condition.
- 13) Verify response time of the network connectivity less than 5 mbps and moderate load conditions.
- 14) Verify response time of the network connectivity less than 5 mbps and normal load conditions.
- 15) Check CPU usage of the application and the database server under peak load conditions
- 16) Verify response time of the Application Under Load is within an acceptable range when the network connectivity is low
- 17) Verify Process count is not more than 4 secs when 1000 users access the number of VM simultaneously.
- 18) Verify Bandwidth usage is not more than 4 secs when 1000 users access the number of VM simultaneously.

- 19) Verify Bandwidth usage can acceptable range when the network connectivity is low
- 20) Check the Error Count that the network connectivity less than 5 mbps.

4.2. Simulation analysis:

As an on-demand service, SOASTA CloudTest is deployed, utilizing the cloud to deliver the load. It is comprised of the above-described approach, the services provided by our professional load testers, and the Global Cloud Test Platform that includes a load generation cross-cloud infrastructure. Open source libraries are a fundamental aspect of the service within the application, used for the provision of different functions in the product. As part of the operation, SOASTA supplies the applications. CloudTest is used to deploy the proposed work and the running conditions along with illustration of generated results are described in GUI model of figure 4 using distributed cloud architecture, complemented by a testing system behind the firewall.

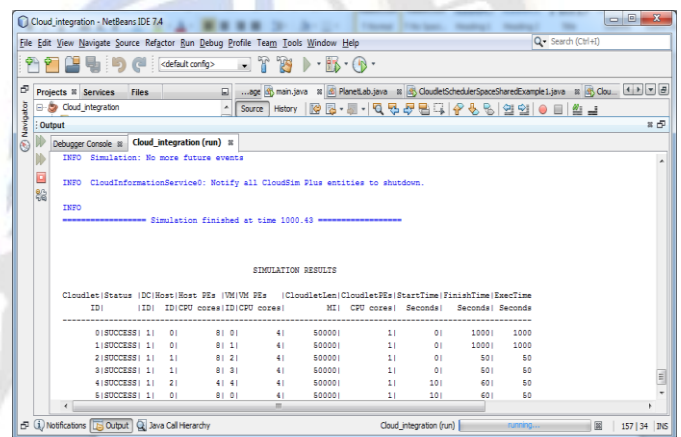
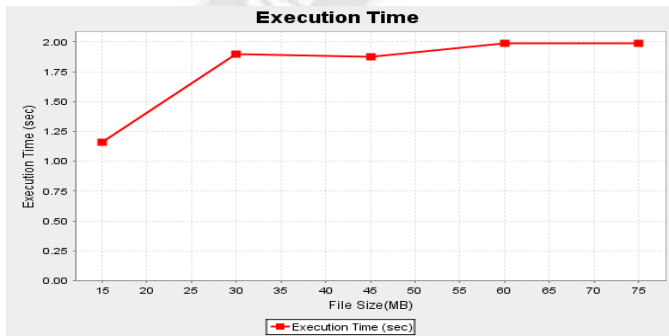


Figure:4 over all Simulation result of the proposed work

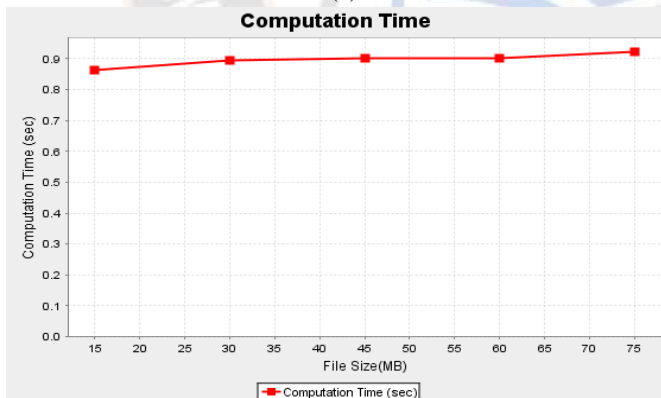
The Global Cloud Test Framework is designed to support additional tools, including Pache JMeter, the most common open source load monitoring tool, while customers can use SOASTA’s application for test development and execution. The SOASTA framework reduces the complexity and time of the cloud deployment of JMeter scripts, making it significantly simpler to build, deploy, conduct and review web-scale load and performance tests for the JMeter group. Without change, JMeter scripts run. SOASTA takes care of managing and provisioning servers and running the test once the test is installed. As a result of the experience of deploying to the cloud, the main features have developed into this method are, Amazon EC2 is the first environment for deployment. Since the load and performance testing criteria matched almost all of the above mentioned characteristics, the implementation of a cloud infrastructure by Amazon is a perfect match.

4.2.1. Performance testing analysis:

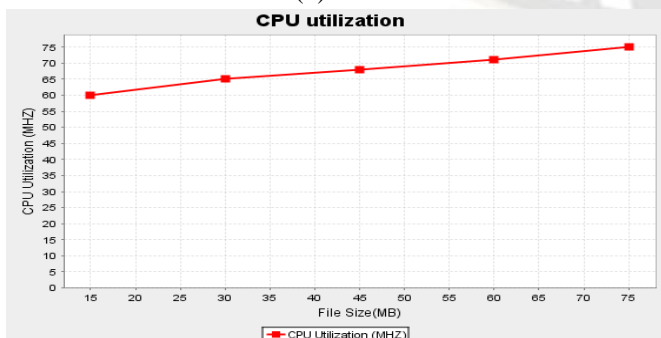
While several published papers in the past two decades discuss system performance testing and scalability assessment, most of them discuss problems and solutions in traditional distributed computing or web-based software systems. Since these systems are set up with pre-configured system resources and infrastructures, performance testing and scalability assessment are typically carried out in a static and pre-fixed system context, so the basic features in cloud testing, such as execution time, computation time, CPU consumption, dynamic scalability, memory utilization, arrival time, waiting time, Cache size, buffer size, scalable testing environments, SLA based requirements, and cost-models are not taken into account in current evaluation metrics, frameworks, and solutions did not consider in prior researches.



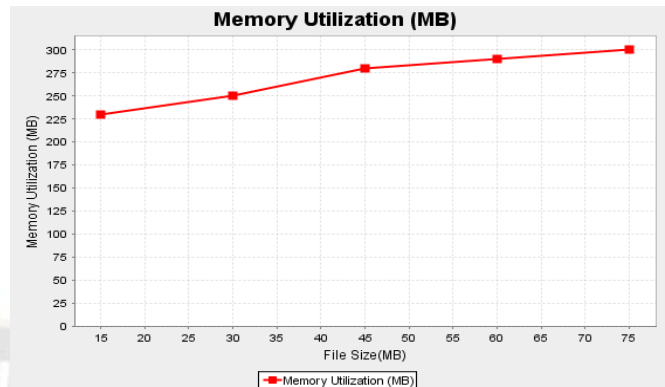
(a)



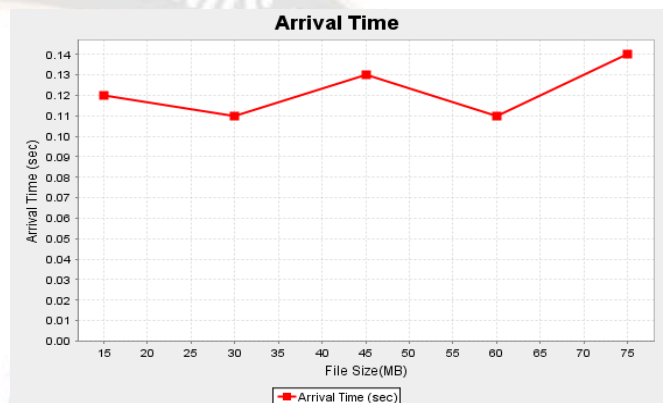
(b)



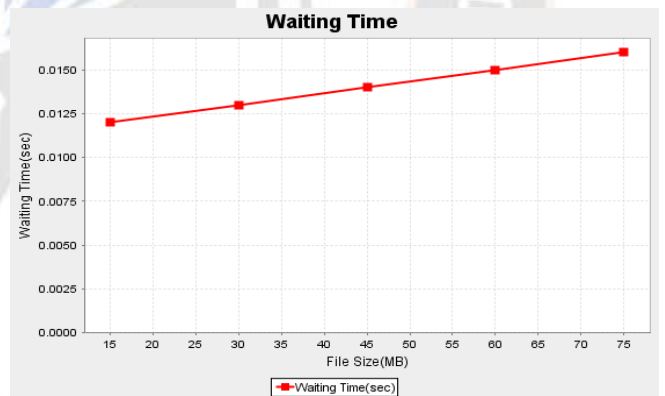
(c)



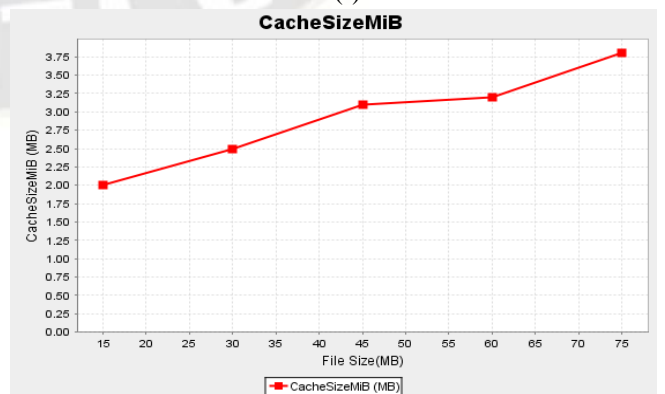
(d)



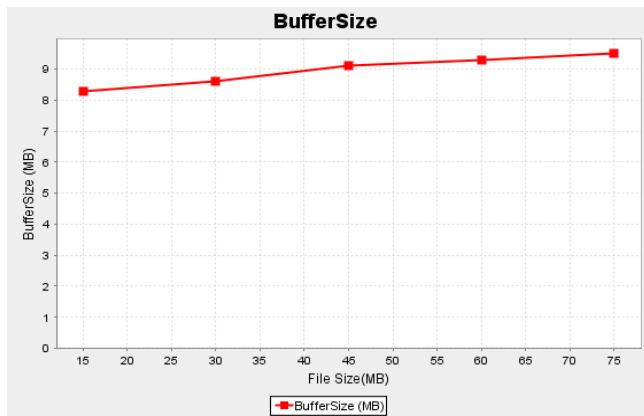
(e)



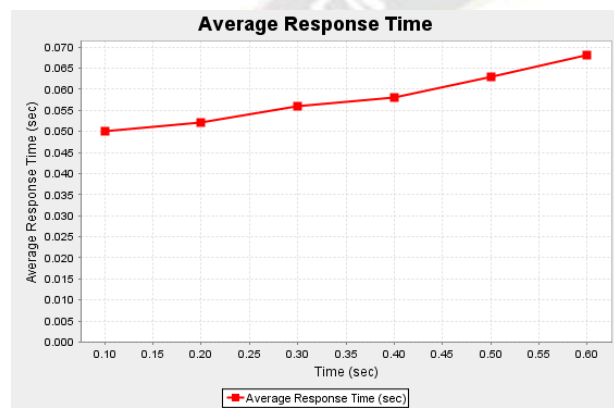
(f)



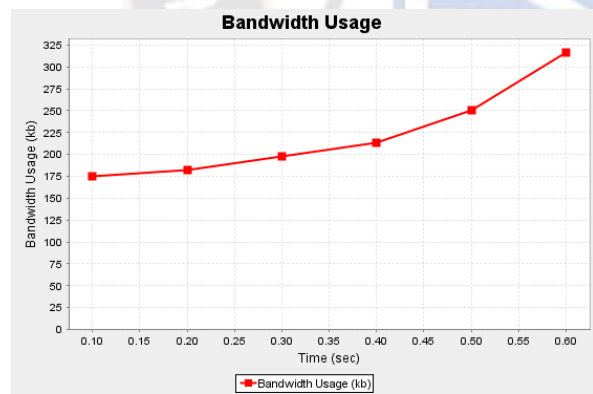
(g)



(h)



(i)



(j)

Figure: 5- proposed framework performance testing analysis; a) execution time b) computation time c) CPU utilization d) memory utilization e) arrival time f) waiting time g) cache size h) buffer size i) average response time j) bandwidth usage

In figure described the overall performance testing analysis for proposed framework, figure 5 a illustrates an execution time performances with file size is 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their execution time is 1.163 sec, 1.896 sec, 1.875 sec, 1.985 sec, and 1.986 sec respectively. Thus it shown the proficiency of the proposed work in cloud testing perform shorten the time interval for execute the program features.

In figure 5 b illustrates the computation time performances for the proposed work with file size is 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their computation time is 0.91 sec, 0.94 sec, 0.95 sec, 0.95 sec, 0.97 sec respectively. Hence, it described the competence of the proposed work in terms of less computation time.

In figure 5 c defined the CPU utilization for the file sizes 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their CPU utilized frequency is 60 MHz, 65 MHz, 68 MHz, 71 MHz, and 75 MHz respectively.

In figure 5 d described the experiment utilization of memory for the file sizes 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their utilized frequency is 230 MB, 250 MB, 280 MB, 290MB, and 300 MB respectively. Less utilization of memory in cloud testing is one of the major considerations hence, it shown the proposed work outperformed the optimization of database configuration.

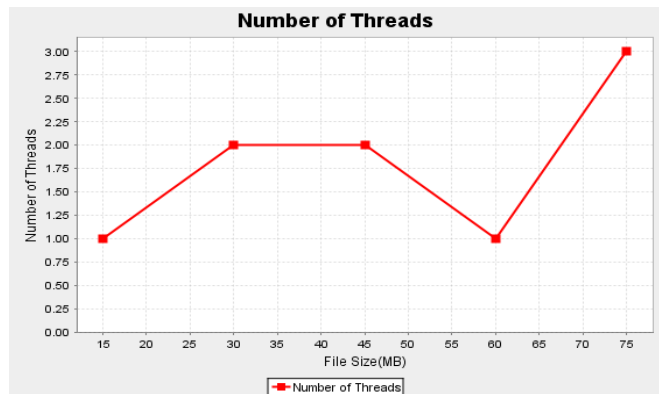
In figure 5 e illustrates arrival time during the transfer of packet when testing, here the experiment contains the file size is 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their corresponding arrival time is 0.12 sec, 0.11 sec, 0.13 sec, 0.11 sec, and 0.14 sec respectively. In figure 5 f shown the waiting time of the various file size is 4 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their corresponding waiting time is 0.012 sec, 0.013 sec, 0.014 sec, 0.015 sec and 0.016 sec respectively. In figure 5 g illustrates the obtained cache sizes during experimentation utilizing file size is 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb, and their corresponding cache size is 2 MB, 2.5 MB, 3.1 MB, 3.2 MB and 3.8 MB respectively. Thus it illustrates the based on proposed methods utilized the extremely less cache sizes.

In figure 5 h, and 5i, illustrates the buffer size, average response time with the file sizes 24 Mb, 30 Mb, 34 Mb, 60Mb, 74 Mb and their buffer size values are 8.3 MB, 8.6 MB, 9.1MB, 9.3 MB, 9.5MB and average response time is 0.05 sec, 0.052 sec, 0.056 sec, 0.058 sec, 0.063 sec, and 0.068 sec respectively.

In figure 5 j described the usage of bandwidth based on various time intervals is 0.1 Sec, 0.2 sec, 0.3 sec, 0.4 sec, 0.5 sec and 0.6 sec and the corresponding bandwidth usage is 175 Kb, 182 Kb, 198 Kb, 213 Kb, 250 Kb, 316 Kb. Then the significant security testing analysis is described in following section.

4.2.2. Security testing analysis:

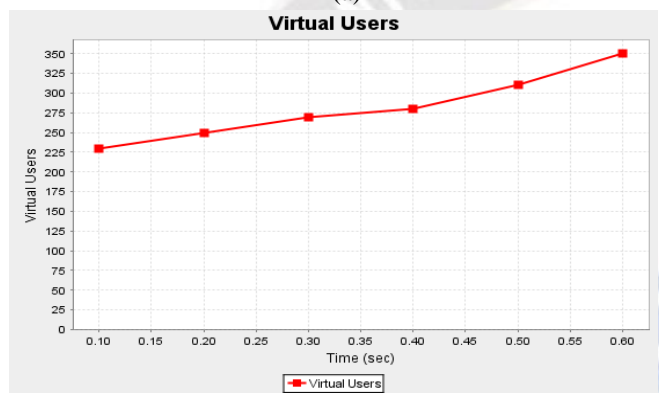
With many open questions in today's software testing culture, security testing has become a hot research topic. Since security is a major concern within clouds and security services are becoming a critical part of modern SaaS and cloud technology, engineers need to address security validation and quality assurance problems and challenges for SaaS and clouds.



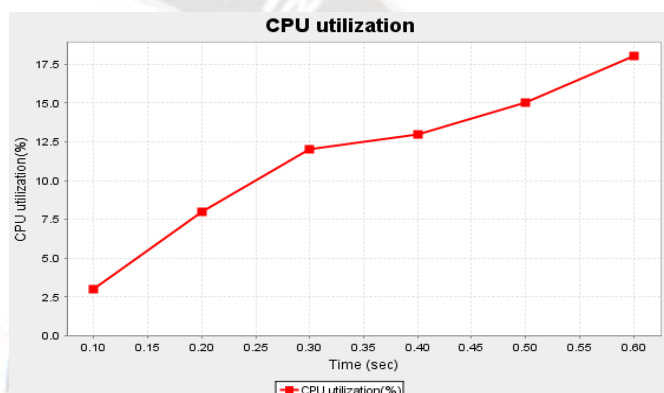
(a)



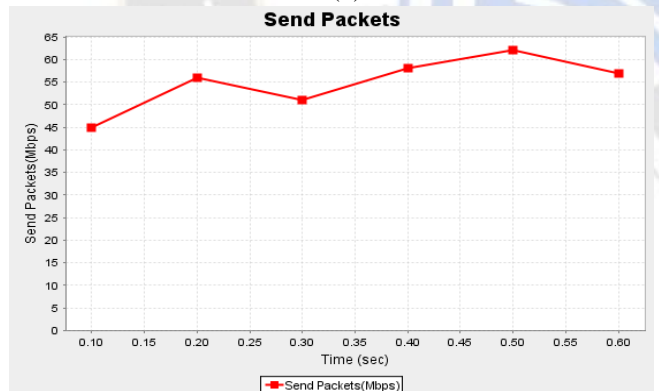
(e)



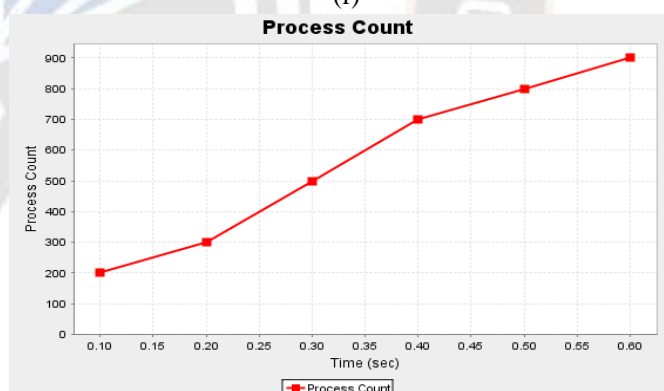
(b)



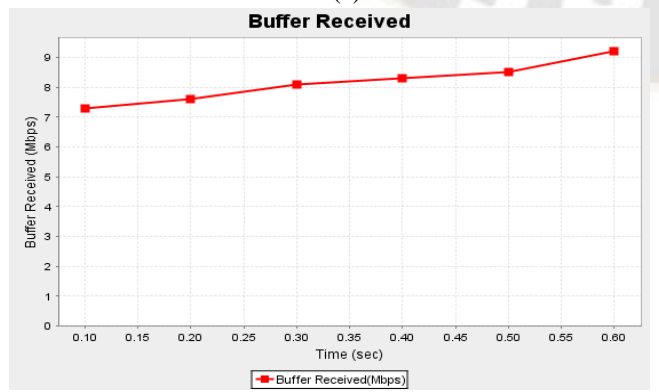
(f)



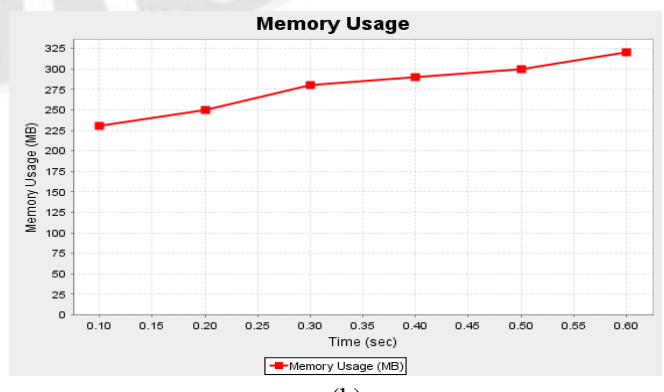
(c)



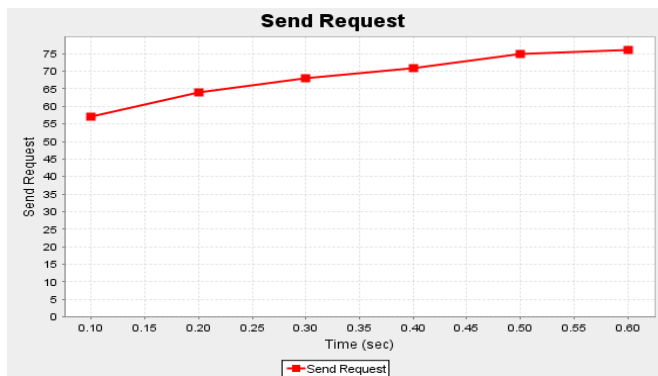
(g)



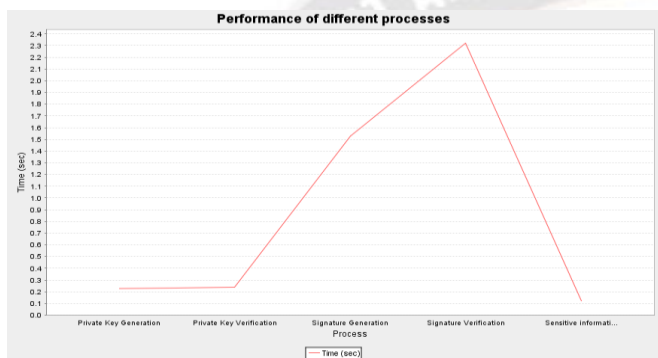
(d)



(h)



(i)



(j)

Figure:6 proposed framework security testing analysis a) number of threads b) virtual user, c) send packets, d) buffer received, e) error count, f) CPU utilization, g) process count h) memory usage, i) send request j) key generation performances

Here the paper accomplished the experimental analysis are the determination of threads number, virtual users, send packets, received buffer, error count, CPU utilization, process count, memory usage, send request, received request, and overall analysis of key generation, which is illustrated in figure 6.

4.3. Comparison analysis:

The proficiency of the proposed work is efficiently compared the performance of signature verification and host sealing with normal virtual network function (VNF) launch time, VNF launch time that is illustrated in figure 7. It can be seen from Figure 7 that the time overhead added by the verification and VNF sealing filters for small-sized VNF images is just a few seconds. This is due to the fact that hash digests over them takes less time to measure. For larger VNF images, however, the time overhead increases because it takes longer to calculate the hash digest. In addition, only when the VNF image size reaches 2 GB does the overhead time become meaningful. As the proposed work assume that 2 GB of image size should be

adequate to compile most of the VNF software files, this result looks satisfactory.

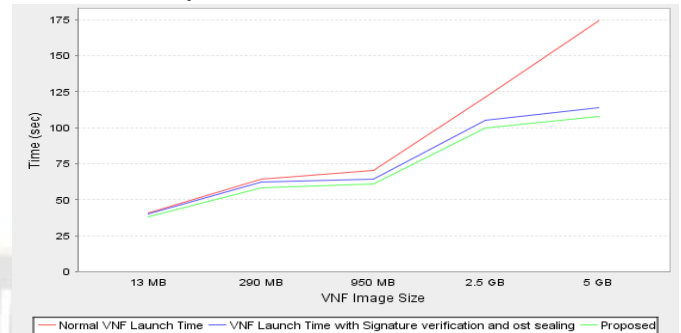


Figure:7 Comparison of signature verification and host sealing with existing work

Next, to calculate the hash digest of a VNF picture, we measured the effect of using different hashing functions. In terms of performance and protection, this paper compare MD5, SHA1 and SHA256 [25] are the Hashing algorithms tested in this experiment. While it is not a feasible choice to use the MD5 algorithm as it is cracked down long ago, the author used it in this paper because it is still commonly used in the verification of software integrity. The time it takes for these three hashing functions to measure the hash digest versus VNF image size is plotted in Figure 8. It can be shown that it takes less time for MD5 and SHA1 algorithms to measure the hash digest, but they are considered insecure. SHA256, on the other hand, is the most stable hashing feature, but it requires greater overhead time. Figure 8 shows that for all three hashing functions for VNF images with a size smaller than 950 MB, the hash digest calculation time is almost the same. The time gap appears to expand for VNF images exceeding 2.5 GB. Nevertheless, as the time overhead is not very large compared to other hashing functions, the proposed work outperform the hash calculation time.

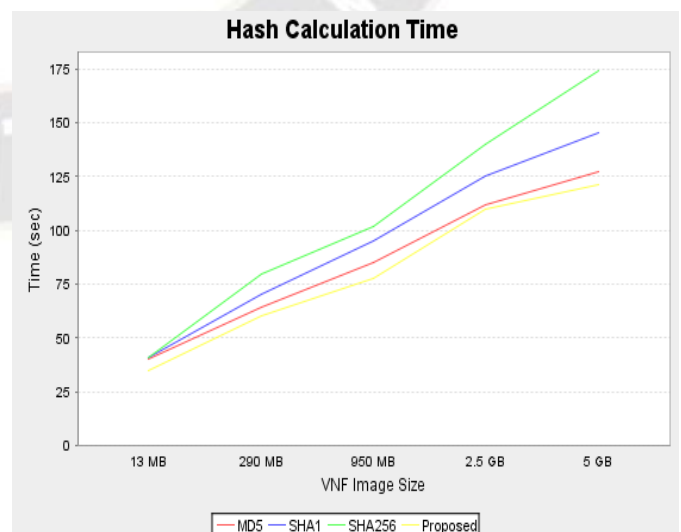


Figure: 8 Comparison of hashing function

The paper installed SecO on two separate platforms in order to test the robustness of the signature verification and VNF-Host sealing method: first in a virtual machine using a KVM hypervisor, and second in a Docker container [26].

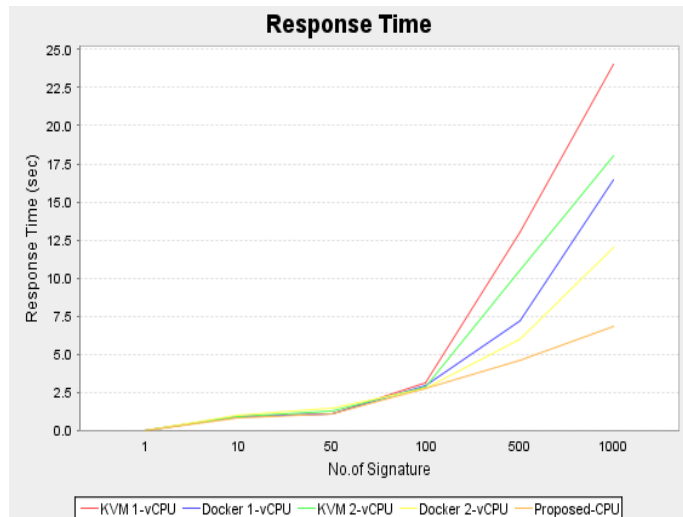


Figure 9- Comparison of response time

Using two separate configurations (i.e. with 1-vCPU and 2-vCPU), we tested SecO's response time. In terms of signature authentication and sealing requests, SecO's output is assessed against its response time for processing them. Figure 9 depicts SecO's response time. It becomes evident that in all four configurations, the time taken by SecO to respond to 100 simultaneous signature verification and sealing requests is much the same. When 500 or more simultaneous requests arrive at SecO, this becomes important. In responding to the verification and sealing requests, SecO containers running on Docker are the fastest of all. That is due to the application containers' improved latency and computational performance compared to VMs.

Hence, from aforementioned concerns that is clearly demonstrates the proposed framework accomplished the cloud testing in terms of resource management, security and database configuration which offers the reduction of time and space complexity in software based cloud testing.

V. CONCLUSION

Amid the development of testing as services and cloud technology, this paper presents the productive framework to tackle the waste of resources, software integrity, and poor optimization of space and time. Initially the framework to measure the performance of lightweight containers with reduction of waste resources through implementation of instinctive calibrates service on containerized orchestration platform to control the calibrate-in/ calibrate-out of containers during work load fluctuation. Then container security is huge

vital after resource management, hence the work incorporates the protection strategy based on single trust platform module that extremely strengthens the authentication throughput from host to running state of container. At last due to the diverse query workloads, the novel work commence with end to end isolated database alteration with attempt-defect manner that overcome the shortcoming caused by regression. The outcome of average relative error is 4.28 % thus it described the potent of the research.

References:

1. AT&T, Building an Enterprise-Focused Cloud Environment, http://about.att.com/innovationblog/enterprise_cloud, Accessed on 27.10.2018 (2017).
2. M. Souppaya, J. Morello and K. Scarfone, NIST Special Publication 800- 190 - Application Container Security Guide, Tech. rep., NIST, Accessed on 820 15.10.2018 (Sep. 2017). doi:10.6028/NIST.SP.800-190.
3. Torkura, K.A., Meinel, C.: Towards cloud-aware vulnerability assessments. In: 2015 11th International Conference on SignalImage Technology & Internet-Based Systems (SITIS), pp. 746– 751. IEEE (2015)
4. Mohamed, B., Youness, K.I., Mohamed, M.: Taking account of trust when adopting cloud computing architecture. In: 2nd International Conference on Cloud Computing Technologies and Applications, pp. 101–106. IEEE (2016)
5. Horvath, A.S., Agrawal, R.: Trust in cloud computing. In: SoutheastCon, pp. 1–8. IEEE (2015)
6. Maheshwari, V., Prasanna, M.: Integrating risk assessment and threat modeling within SDLC process. In: International Conference on Inventive Computation Technologies (ICICT), vol. 1, pp. 1–5. IEEE (2016)
7. Shenoy, S., Kuo, T.-T., Gabriel, R., McAuley, J., Hsu, C.-N.: Deduplication in a massive clinical note dataset, pp. 5–16. University of California, San Diego, La Jolla, CA (2017)
8. Vijayakumar, K., Arun, C.: Analysis and selection of risk assessment frameworks for cloud based enterprise applications. In: Special Issue on Biomed Research India - Artificial Intelligent Techniques for Bio-Medical Signal Processing, pp. 1-8 (2017)
9. Vijayakumar, K., Arun, C.: Automated risk identification using NLP in cloud based development environments. J. Ambient Intell. Humaniz. Comput (2017). doi:10.1007/s12652-017-0503-7
10. Ali, M.M., Huda, S., Abawajy, J., Alyahya, S., Al-Dossari, H., Yearwood, J.: A parallel framework for

- software defect detection and metric selection on cloud computing. *Cluster Comput.* 1–15 (2017)
11. Xu, X., Chen, Y., Calero, J.M.A.: Distributed decentralized collaborative monitoring architecture for cloud infrastructures. *Cluster Comput.* 20(3), 2451–2463 (2017)
 12. Amro Al-Said Ahmad, Pearl Brereton, and Peter Andras. 2017. A systematic mapping study of empirical studies on software cloud testing methods. In *Proceedings of the IEEE International Conference on Software Quality, Reliability and Security*. IEEE, 555–562.
 13. Michel Albonico, Amine Benelallam, Jean-Marie Mottu, and Gerson Sunyé. 2016. A DSL-based approach for elasticity testing of cloud systems. In *Proceedings of the International Workshop on Domain-Specific Modeling (DSM'16)*. 8–14
 14. Michel Albonico, Stefano Di Alesio, Jean-Marie Mottu, Sagar Sen, and Gerson Sunyé. 2017. Generating test sequences to assess the performance of elastic cloud-based systems. In *Proceedings of the International Conference on Cloud Computing*. IEEE, 383–390.
 15. Amira Ali and Nagwa Badr. 2016. Performance testing as a service for web applications. In *Proceedings of the 7th IEEE International Conference on Intelligent Computing and Information Systems (ICICIS'15)*. 356–361.
 16. Marco Anisetti, Claudio Ardagna, Ernesto Damiani, and Filippo Gaudenzi. 2017. A semi-automatic and trustworthy scheme for continuous cloud service certification. *IEEE Trans. Serv. Comput.* (2017). <https://doi.org/10.1109/TSC.2017.2657505>. Early Access
 17. Sunitha Badanahatti and Yelisetty Satya Sree Rama Murthy. 2016. Optimal test case prioritization in cloud based regression testing with aid of KFCM. *International Journal of Intelligent Engineering and Systems* 10, 2 (2016), 96–106
 18. Morán, J., Bertolino, A., de la Riva, C. and Tuyá, J., 2018. Automatic testing of design faults in mapreduce applications. *IEEE Transactions on Reliability*, 67(3), pp.717-732.
 19. Hierons, R.M., Merayo, M.G. and Núñez, M., 2018. Bounded reordering in the distributed test architecture. *IEEE Transactions on Reliability*, 67(2), pp.522-537.
 20. Jin, H., Li, Z., Zou, D. and Yuan, B., 2019. Dseom: A framework for dynamic security evaluation and optimization of MTD in container-based cloud. *IEEE Transactions on Dependable and Secure Computing*.
 21. Yao, J., Shoja, B.M. and Tabrizi, N., 2019, June. An Overview of Cloud Computing Testing Research. In *International Conference on Cloud Computing* (pp. 303-313). Springer, Cham.
 22. Nurul, M. and Quadri, S.M.K., 2019, January. Software Testing Approach for Cloud Applications (STACA)–Methodology, Techniques & Tools. In *2019 9th International Conference on Cloud Computing, Data Science & Engineering (Confluence)* (pp. 19-25). IEEE.
 23. Ebadi, Y. and Jafari Navimipour, N., 2019. An energy-aware method for data replication in the cloud environments using a Tabu search and particle swarm optimization algorithm. *Concurrency and Computation: Practice and Experience*, 31(1), p.e4757.
 24. Patil, R., Dudeja, H. and Modi, C., 2019. Designing an efficient security framework for detecting intrusions in virtual network of cloud computing. *Computers & Security*, 85, pp.402-422.
 25. Lal, S., Ravidas, S., Oliver, I. and Taleb, T., 2017, May. Assuring virtual network function image integrity and host sealing in Telco cloude. In *2017 IEEE International Conference on Communications (ICC)* (pp. 1-6). IEEE.